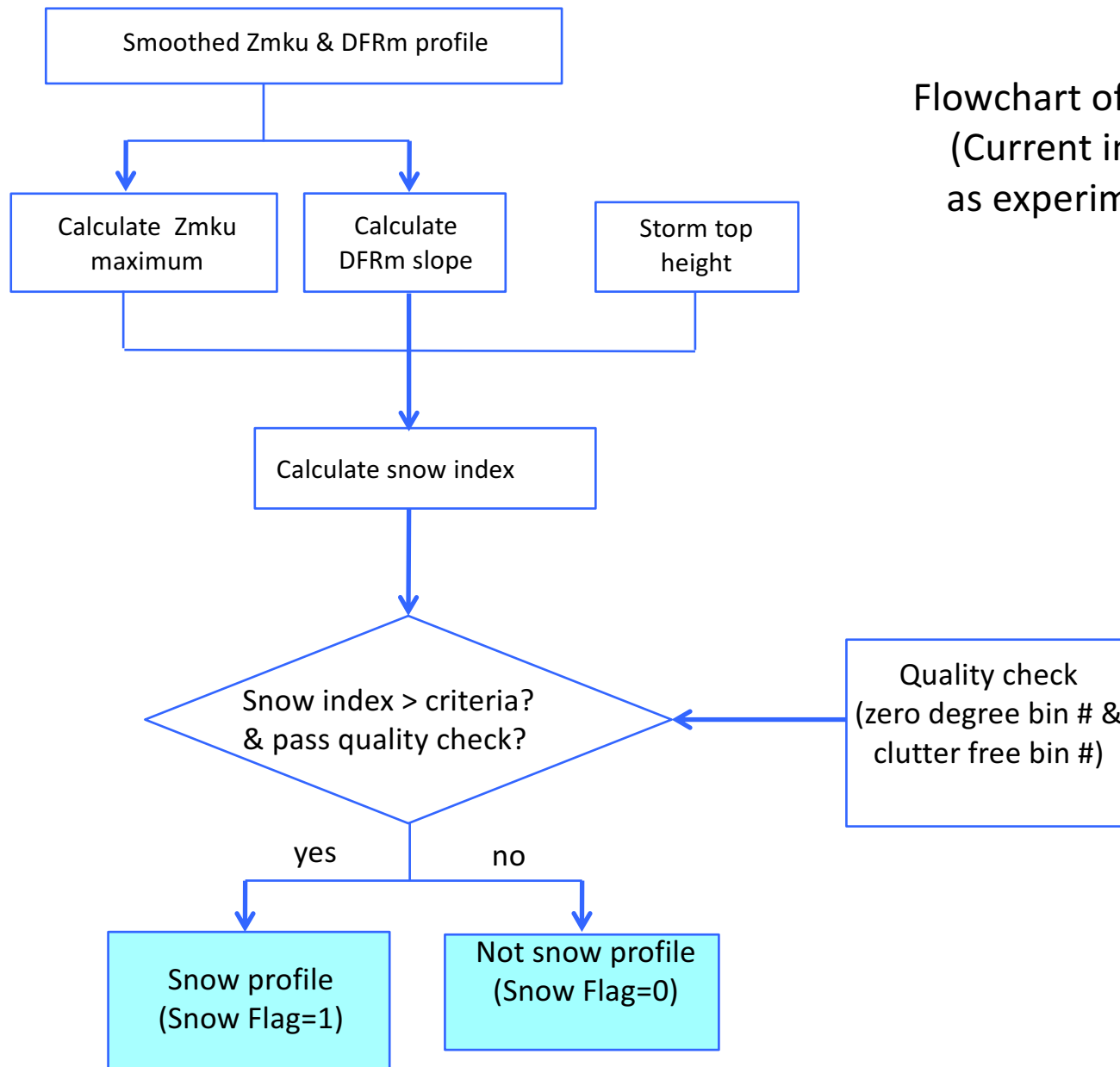


# **Surface snowfall identification using dual-frequency profiles**

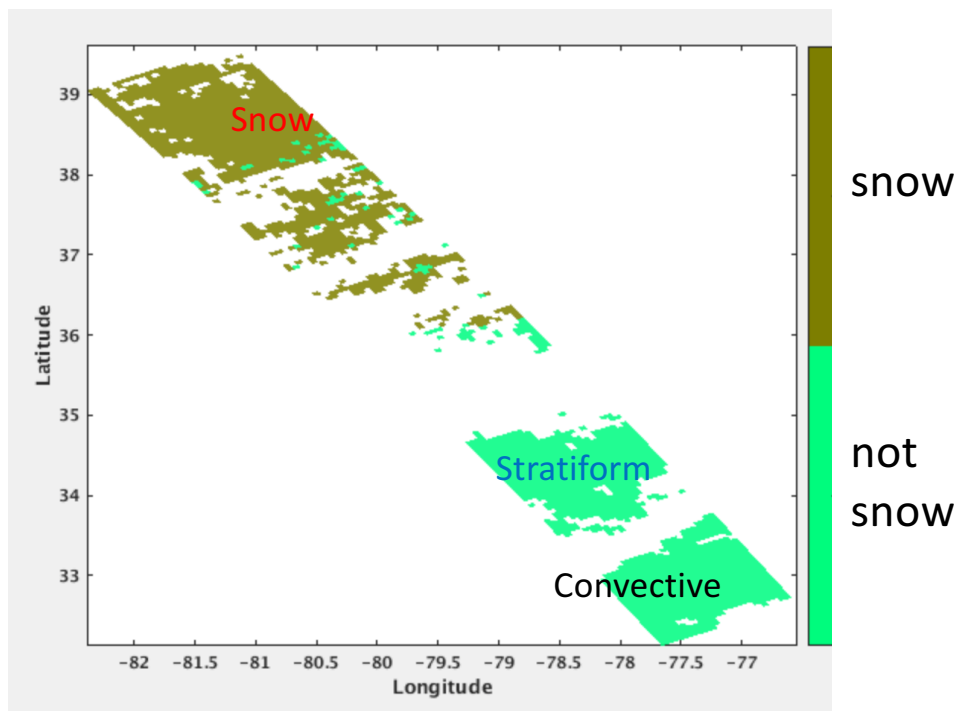
**Chandra and Minda Le**

**CSU**

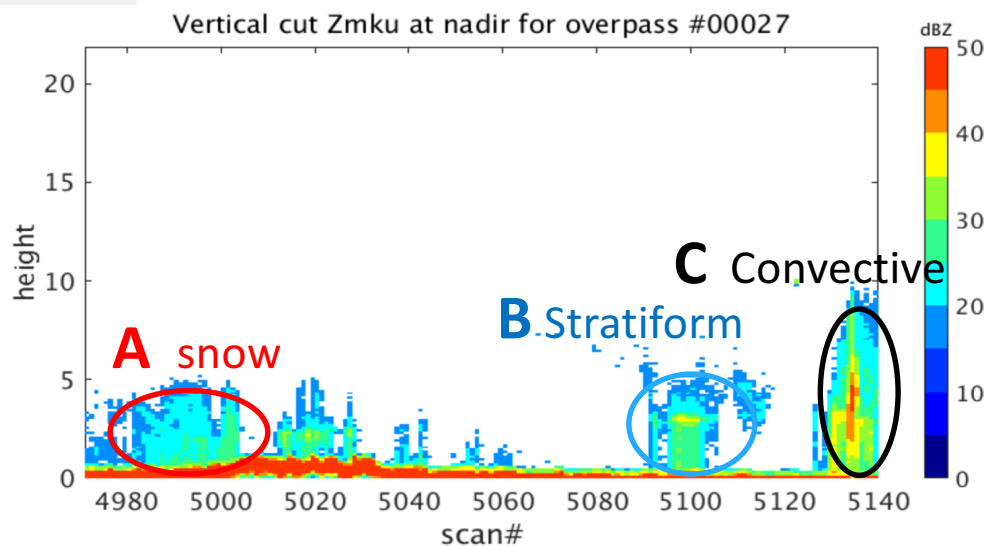
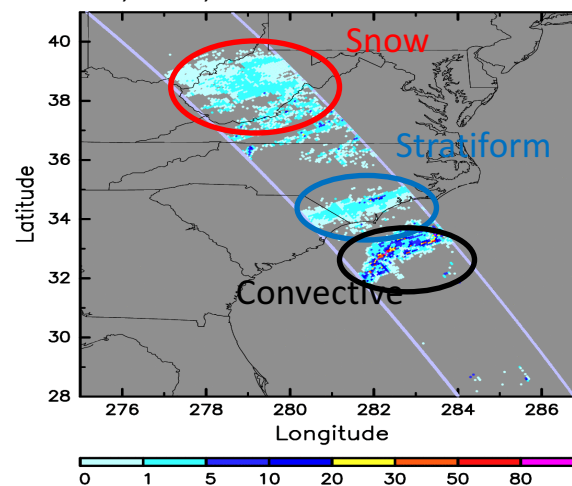
# Surface snowfall detection algorithm using snow index



# Sample performance



Rain rate from Ku on GPM/DPR  
March 17, 2014, orbit 000272 [mm/h]



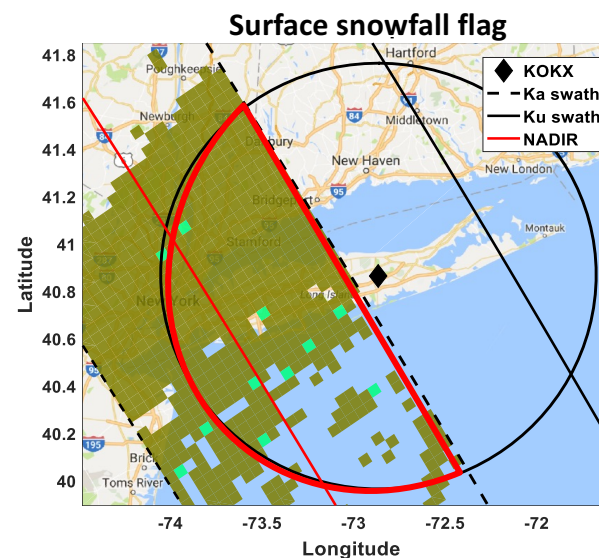
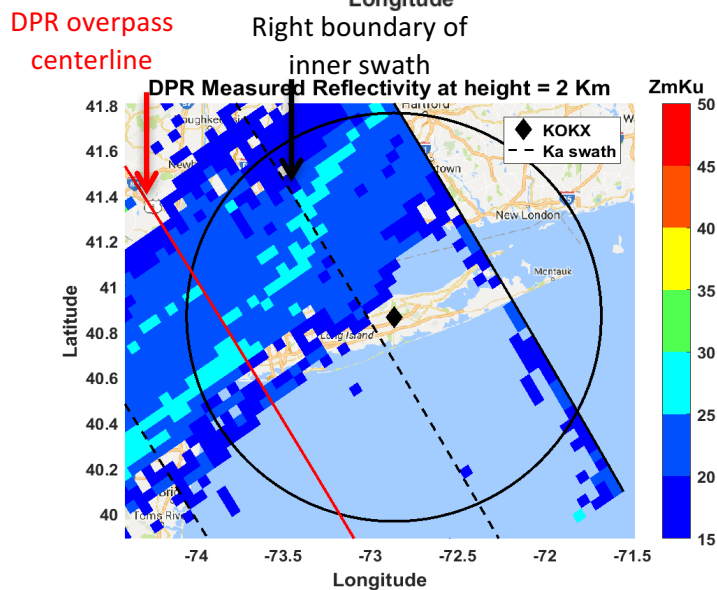
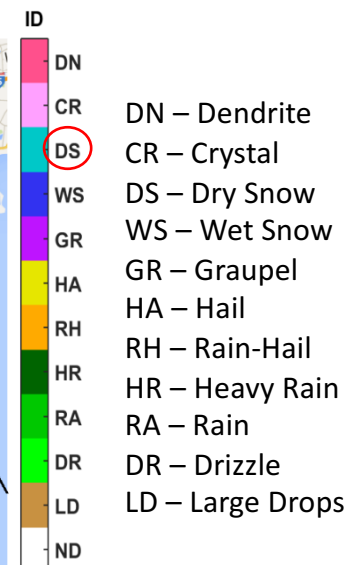
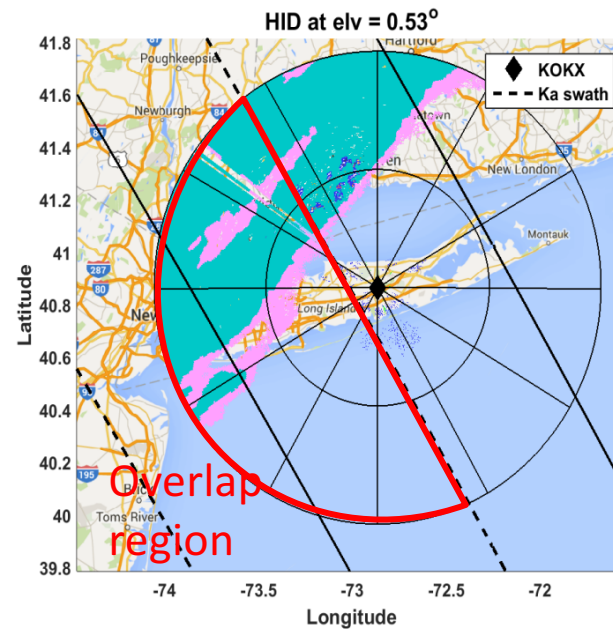
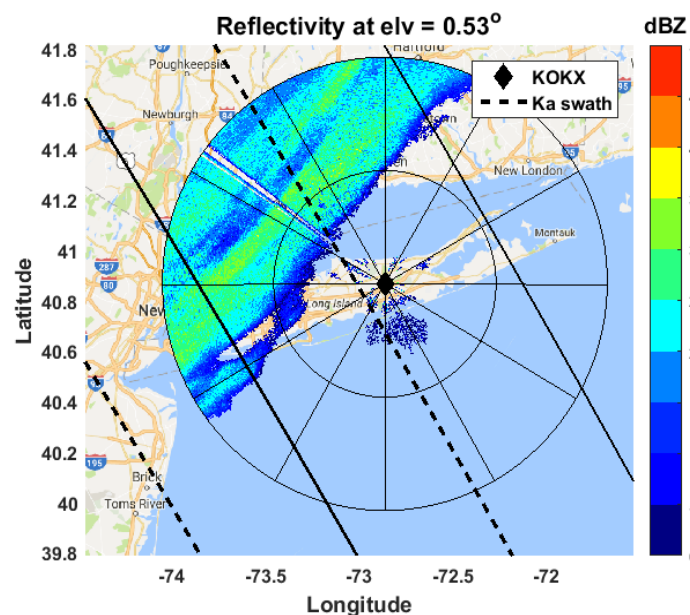
# Ground validation sample case: GPM overpass #4914 with KOKX Radar at Upton, NY

Date: Jan 09, 2015 Time: 12:24:55 UTC



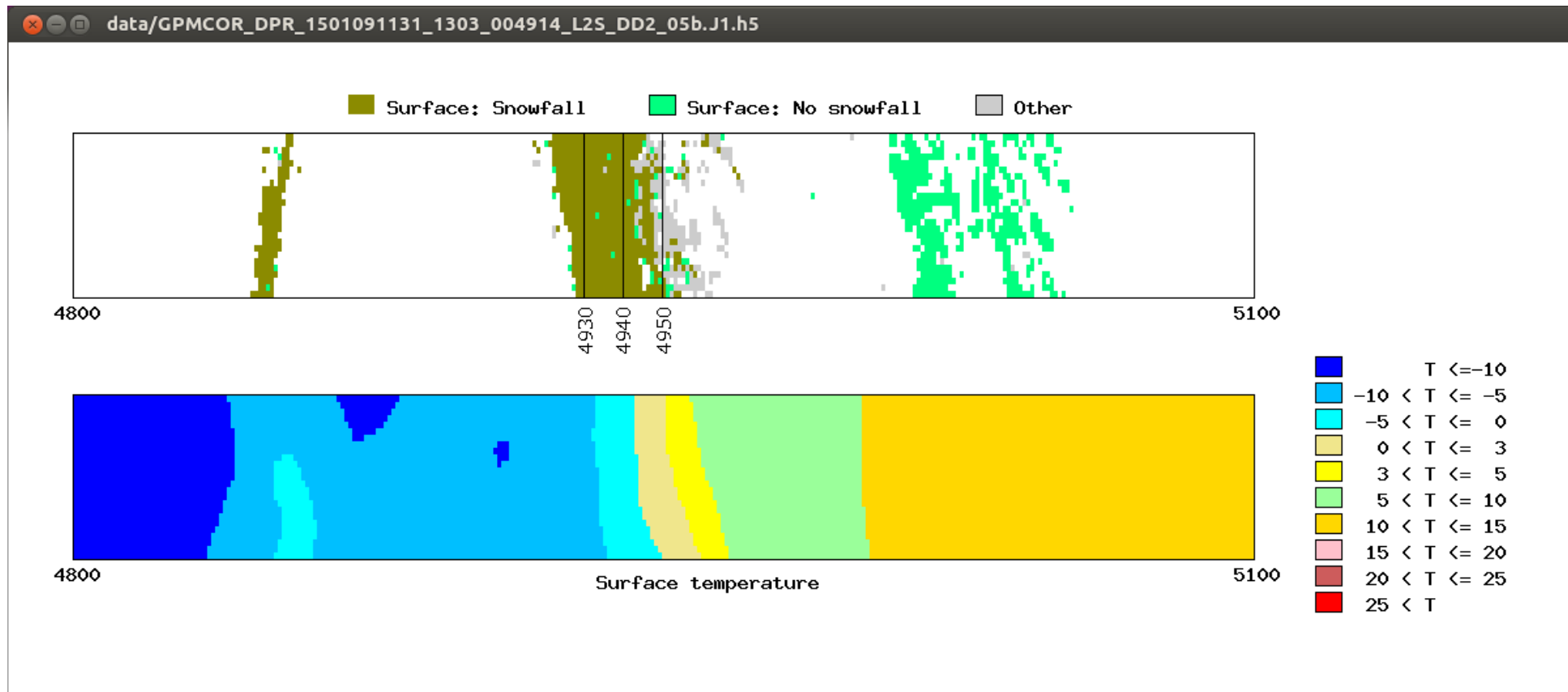


# Hydrometeor class from ground radar matches well with surface snowfall identification method



snow

not snow



Alternate Verification: Orbit # 4914

# Path Attenuation Estimates for the DPR

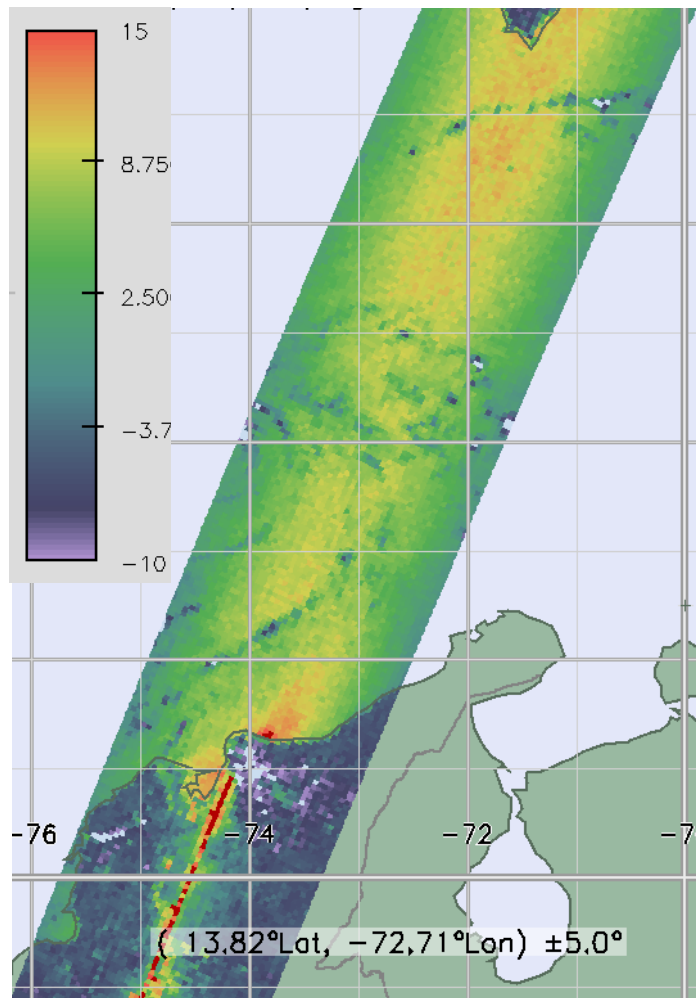
Robert Meneghini<sup>1</sup>, Hyokyung Kim<sup>2</sup>, Liang Liao<sup>2</sup>

1. NASA/GSFC, Code 612

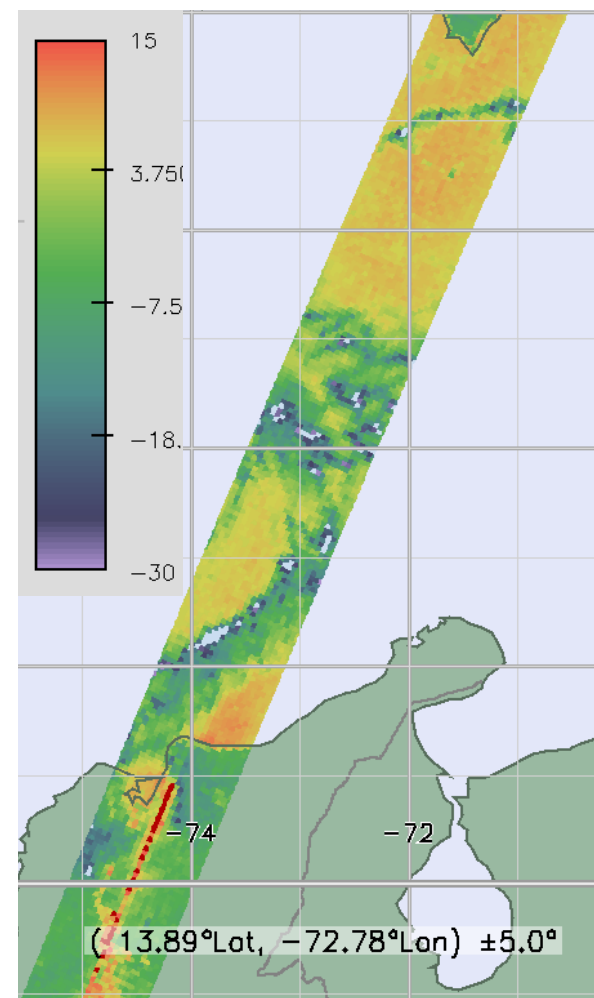
2. Morgan State University/GSFC, Code 612

# Outline

- Background & Status
- Reduction in variability of reference data
- NUBF Mitigation

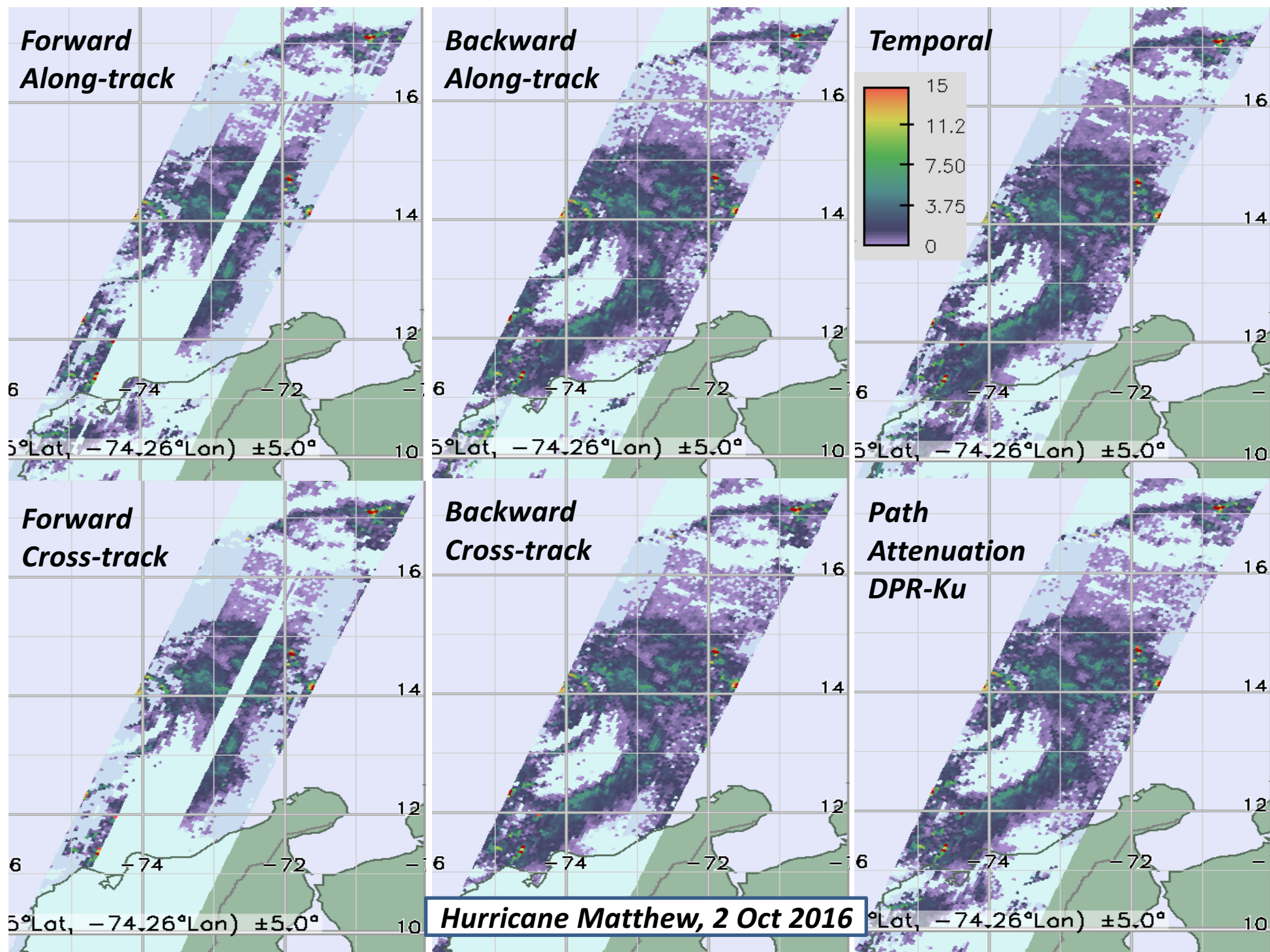


$\sigma^0(\text{Ku})$

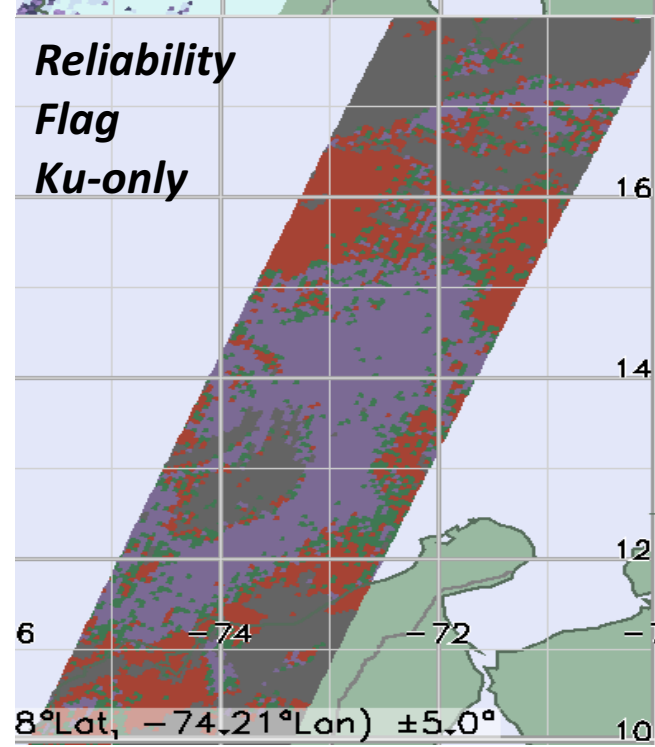
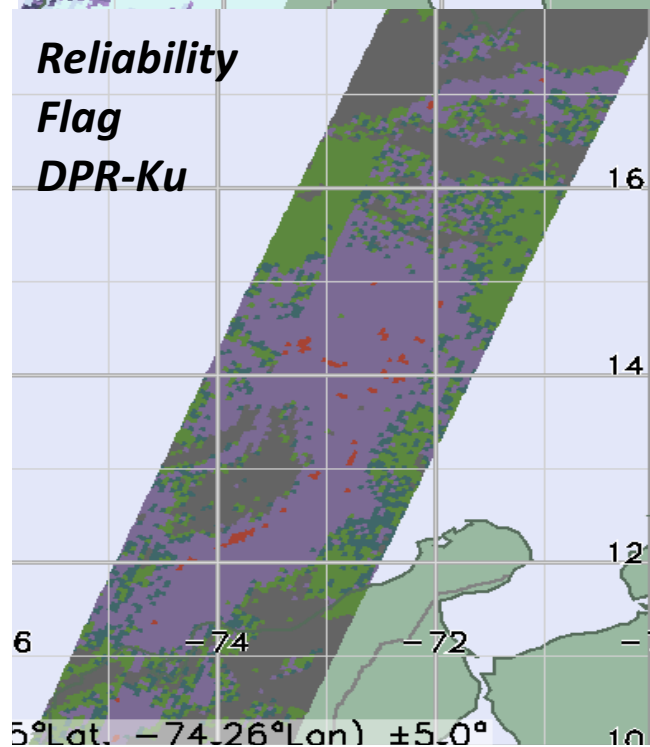
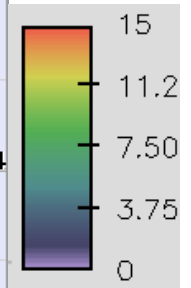
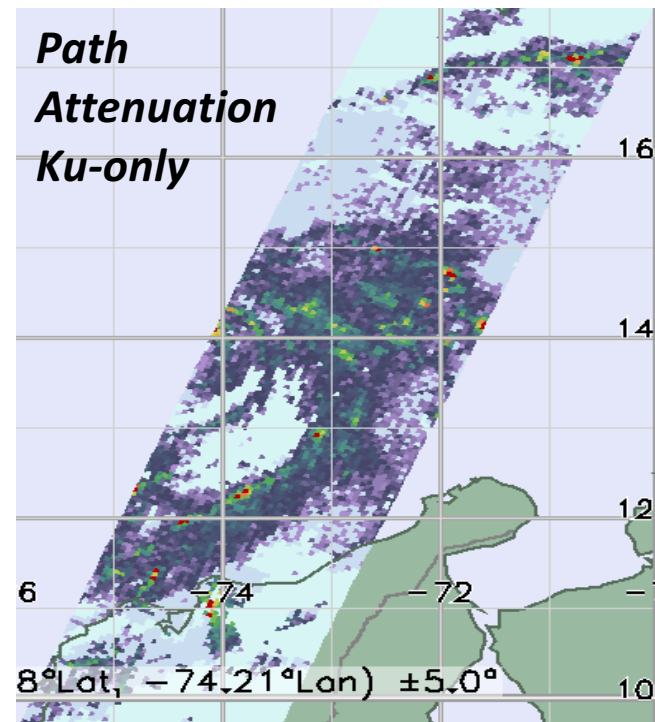
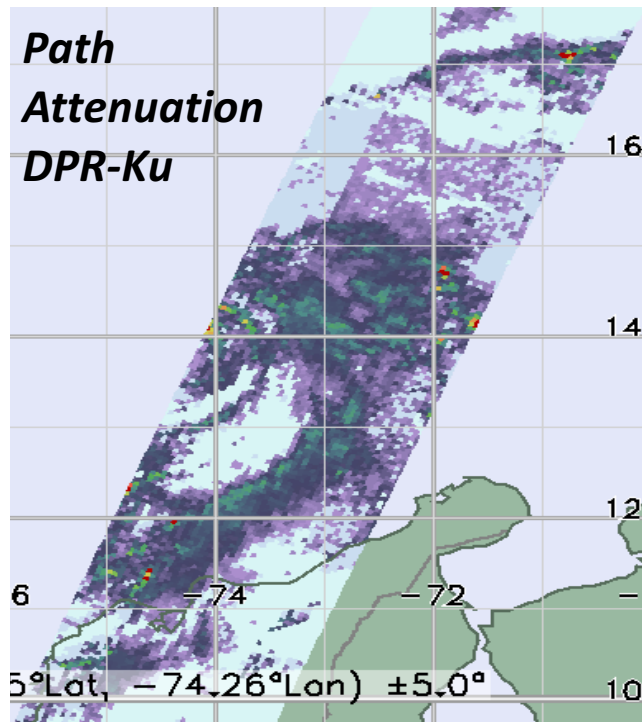


$\sigma^0(\text{Ka})$

***Hurricane Matthew, 2 Oct 2016***





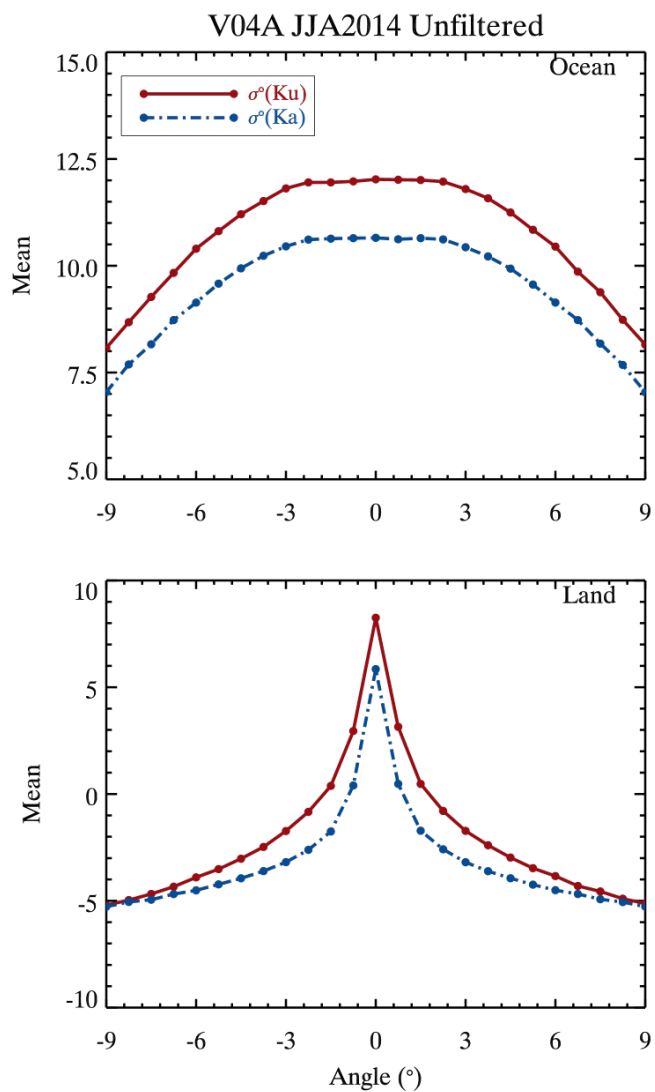


# Status of Algorithm

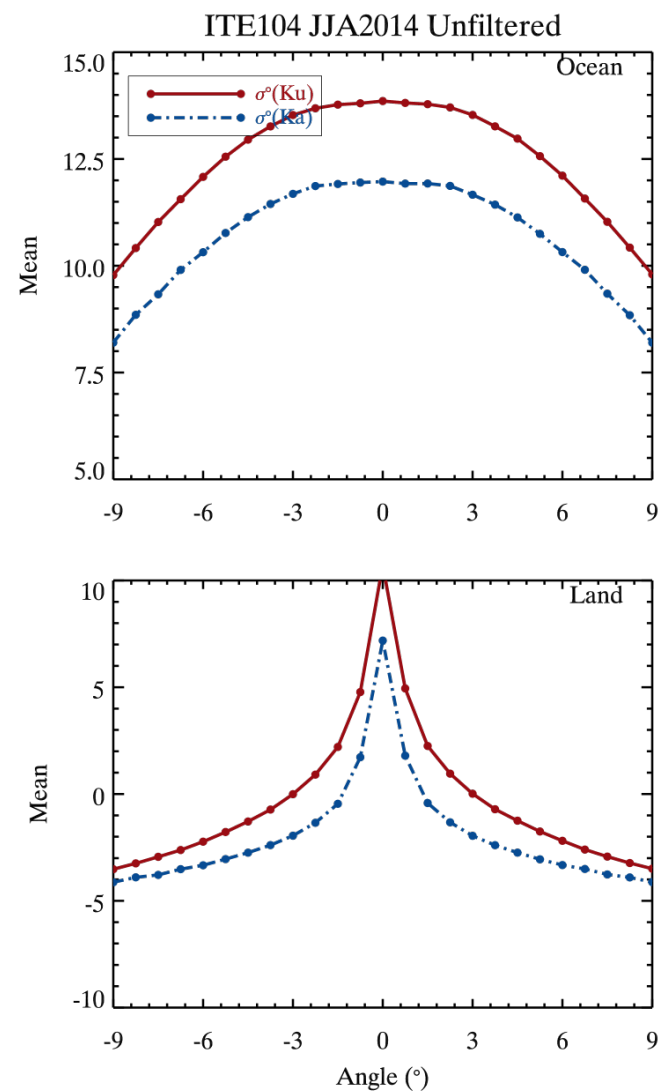
- A number of parameters used to compute  $\sigma^0$  have changed
  - Redo temporal reference look-up tables
  - Compare path attenuations from V4 and V5
- Modify code & reference data for 5 surface types (ocean/land/coast/sea-ice/snow)
- Optional use of corrections for saturated Ku-band  $\sigma^0$  data



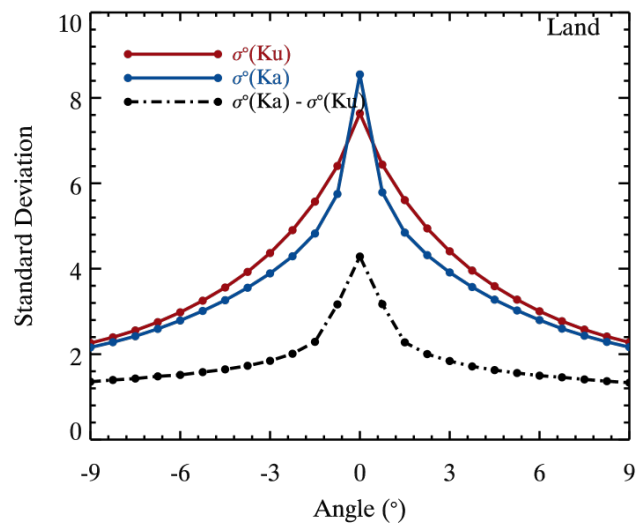
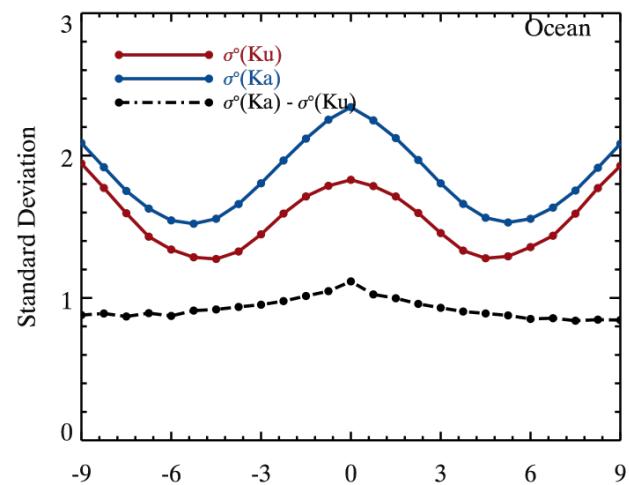
# V4



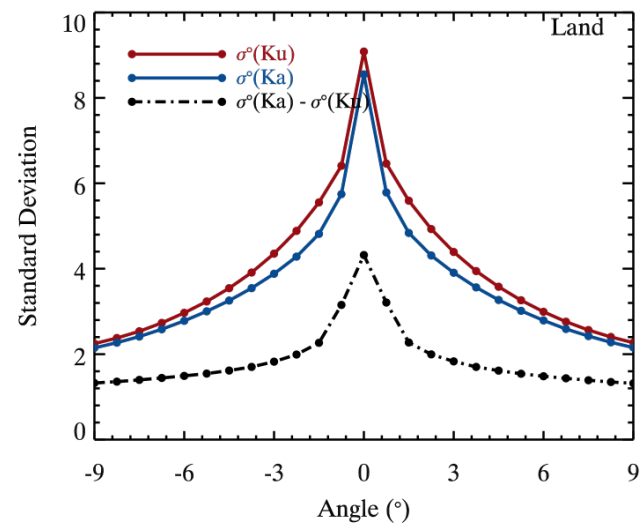
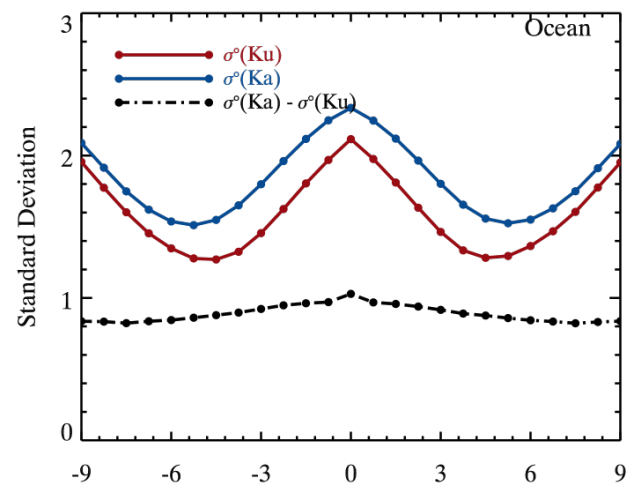
# V5



# V4

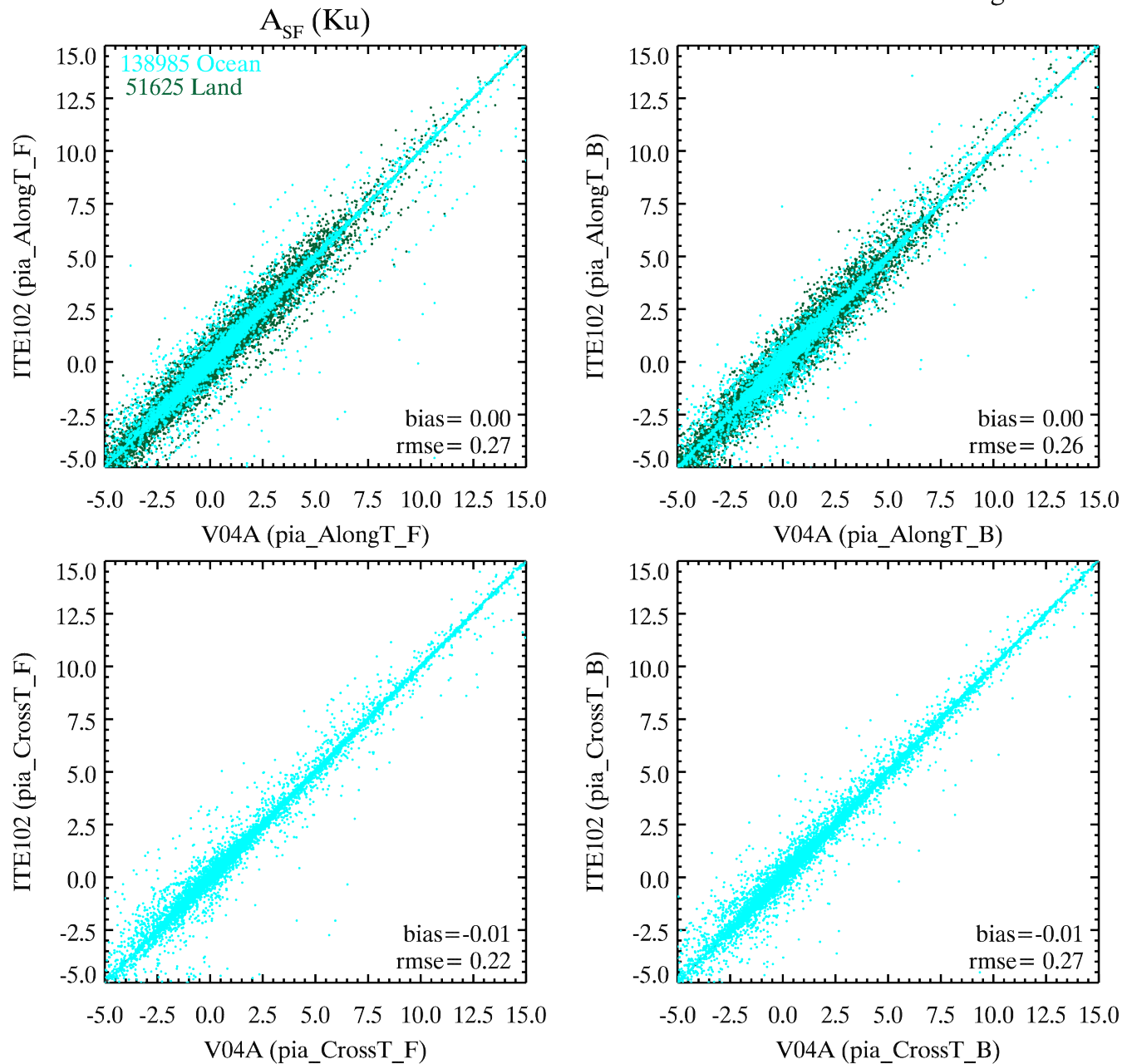


# V5



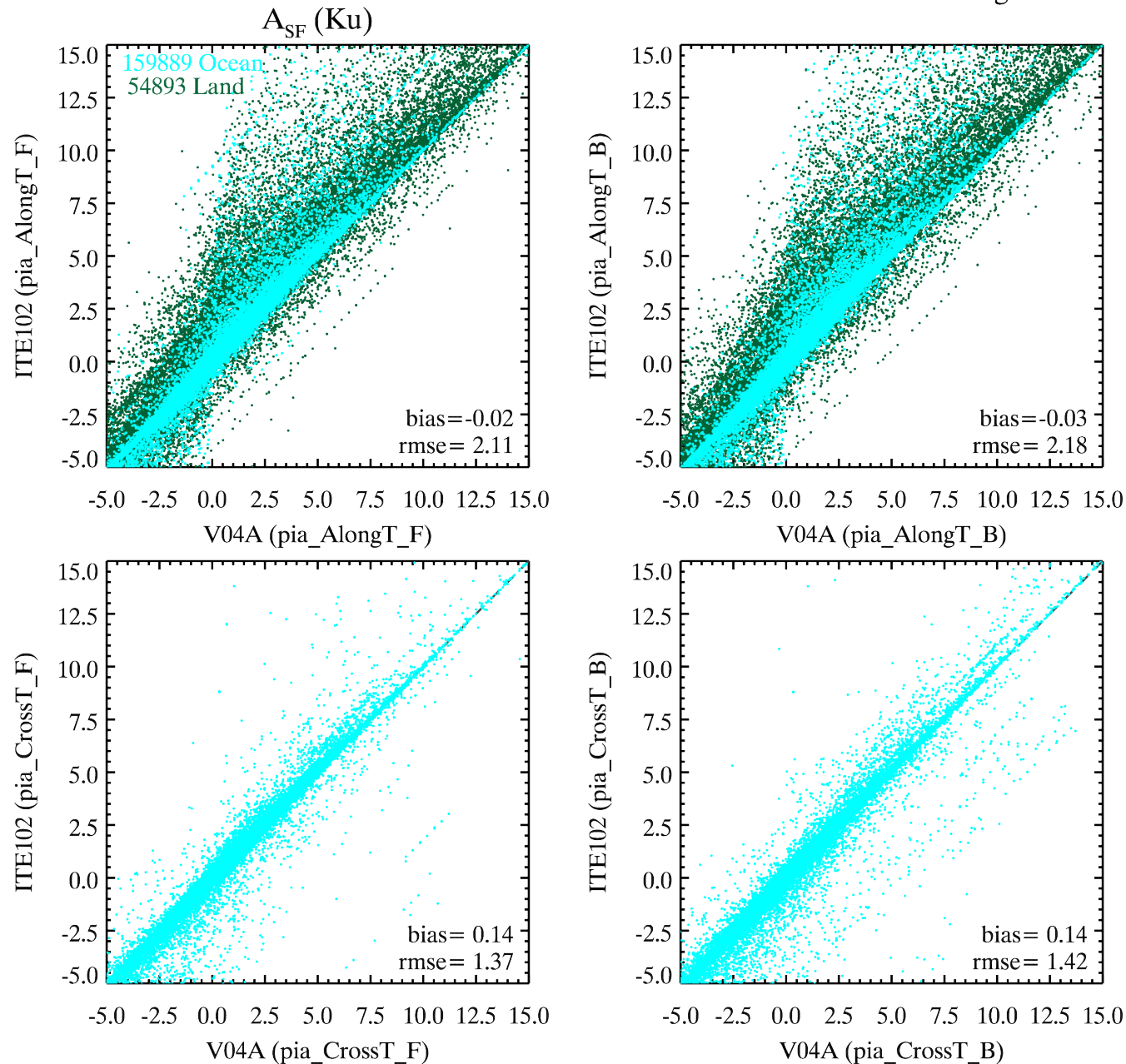
# Path Attenuations: V5 (ordinate) vs V4 (abscissa) [Ku-band]

Angle: 6.75°



# Path Attenuations: V5 (ordinate) vs V4 (abscissa) [Ku-band]

Angle: 0.00°



# How can we improve the estimates of path attenuation?

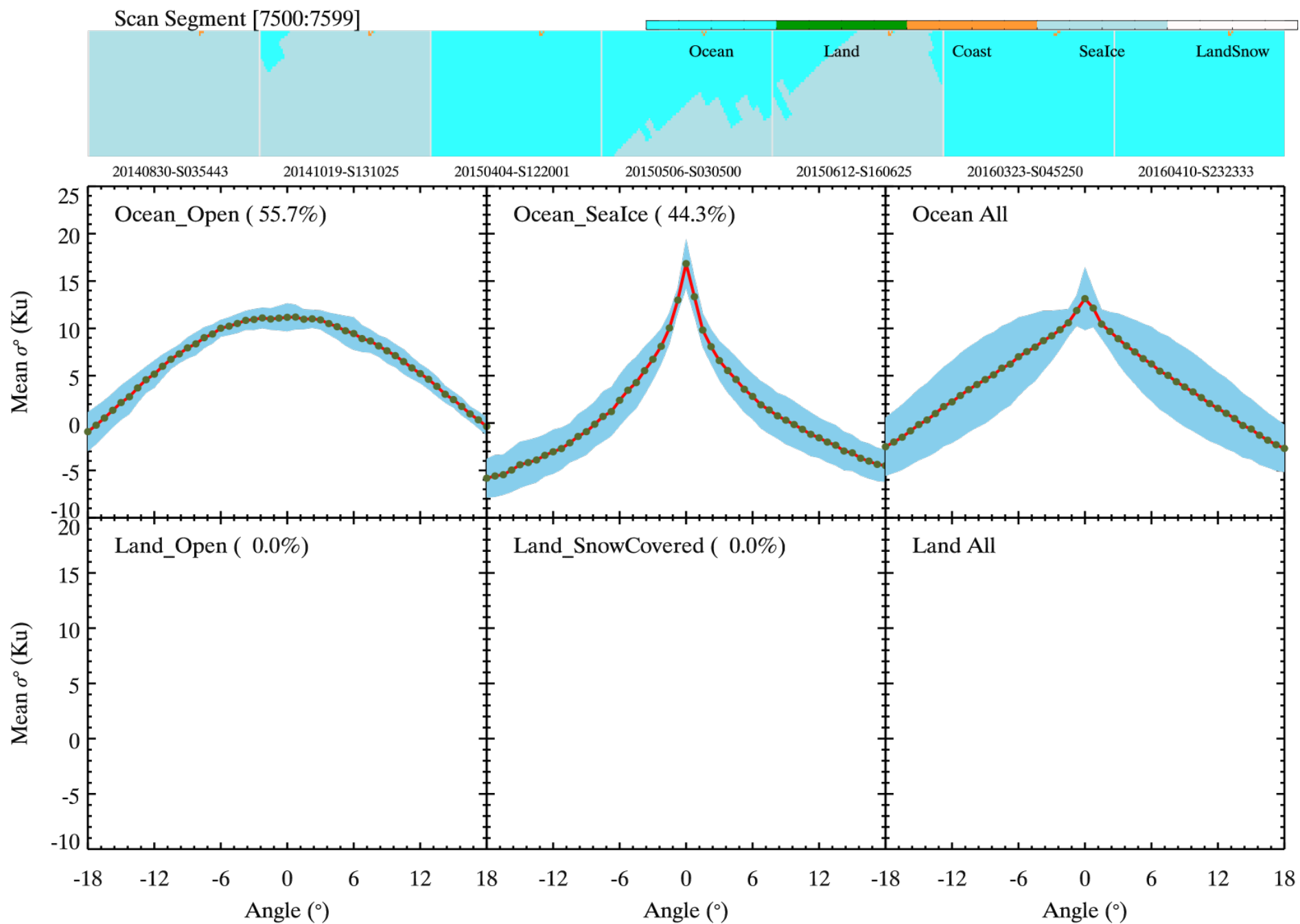
- Reduce the standard deviation of the rain-free  $\sigma^0$  reference data
  - Address the under-sampling problem at near-nadir incidence
  - Reduce the variability of the temporal reference data
  - Add new surface types: sea-ice, snow-covered land
- Reduce the errors caused by non-uniform beam-filling

# Sea-Ice & Snow-covered Land

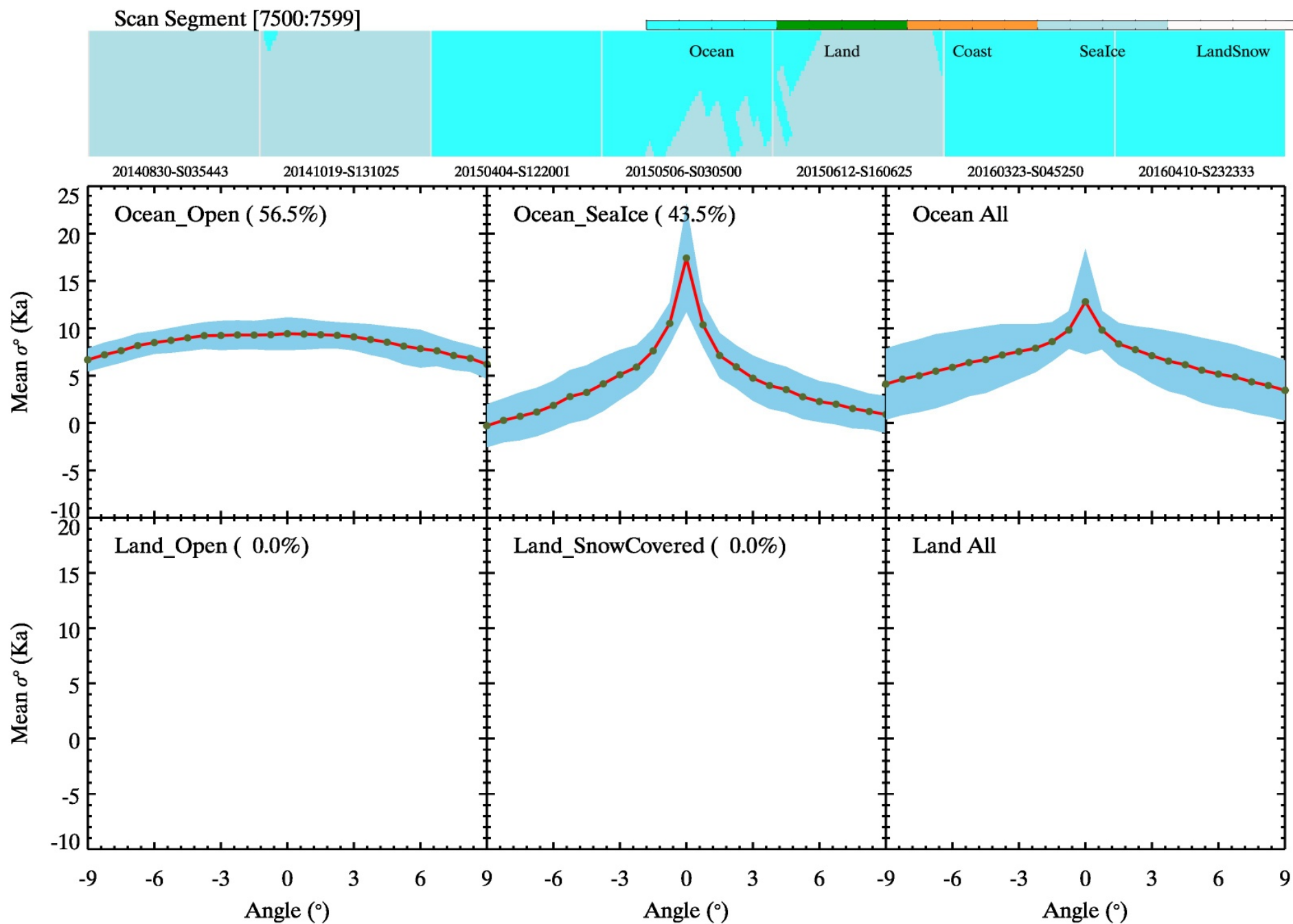
- Motivation
  - The variance of  $\sigma^0$  at high latitudes can be reduced substantially if open ocean and sea-ice cases are separated
    - The mean rain-free  $\sigma^0$  for the combined data generally will be incorrect for both categories
  - Evidence points to some reduction in variance if land & snow-covered land are also distinguished

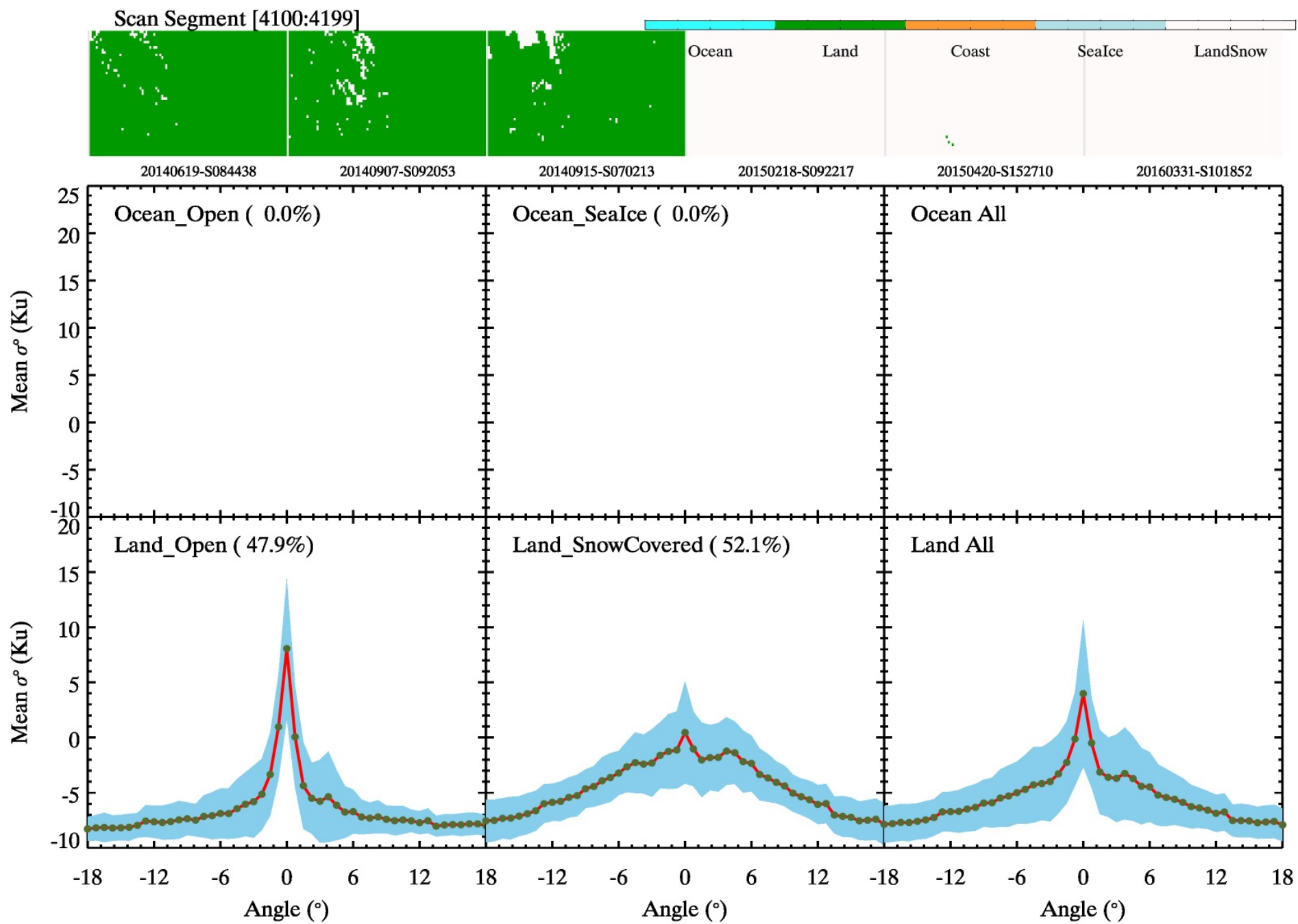
# Orbit Clustering

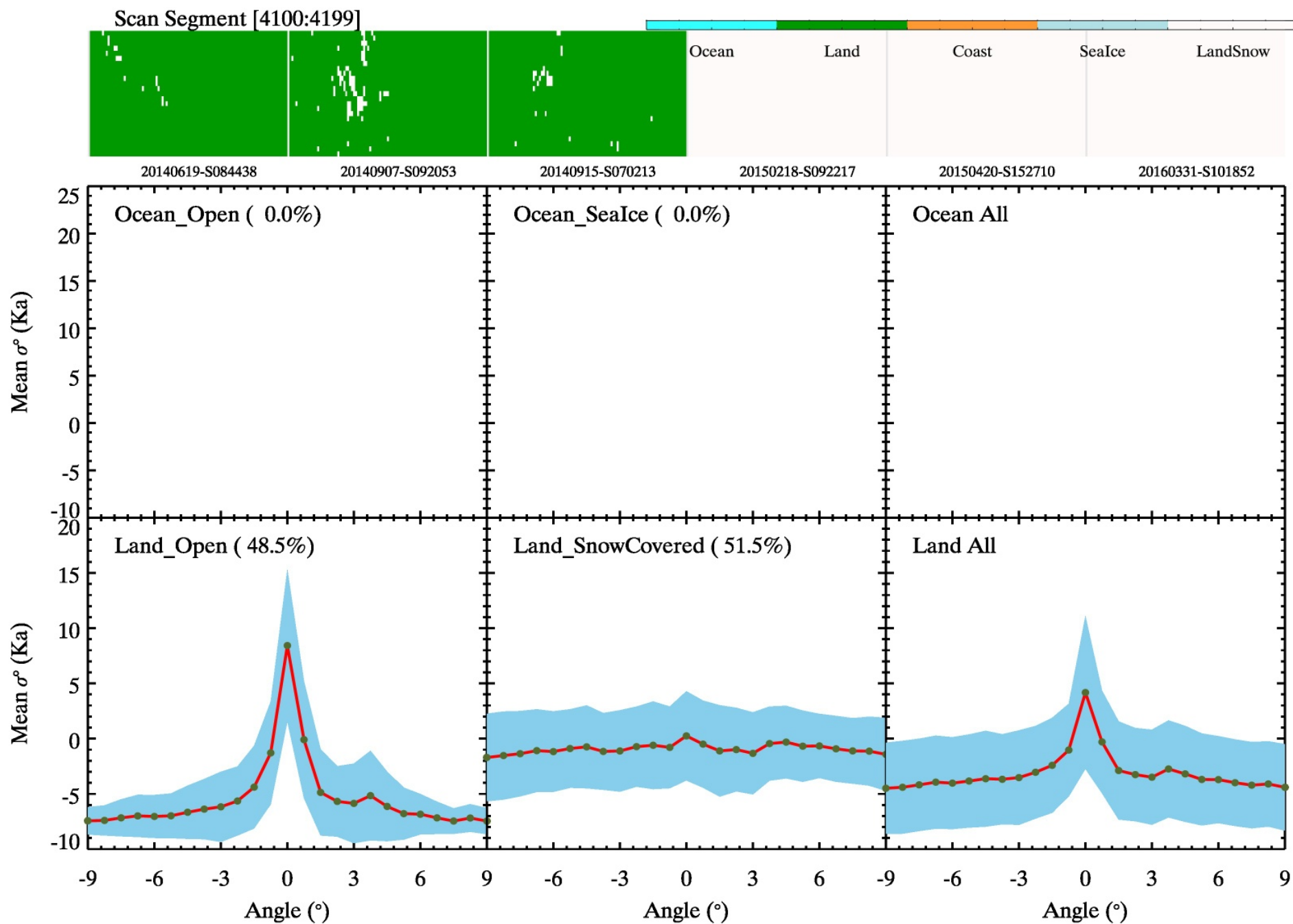
- Insight into the influence of surface type (and accuracy of temporal reference) and can be obtained by looking at tight clusters of orbits
  - Search for sets of orbits (7) with nearly identical trajectories (deviation less than 1 FOV)
  - Data also can be used to test the performance of the temporal reference PIA estimates under ideal conditions (minimum spatial variation)





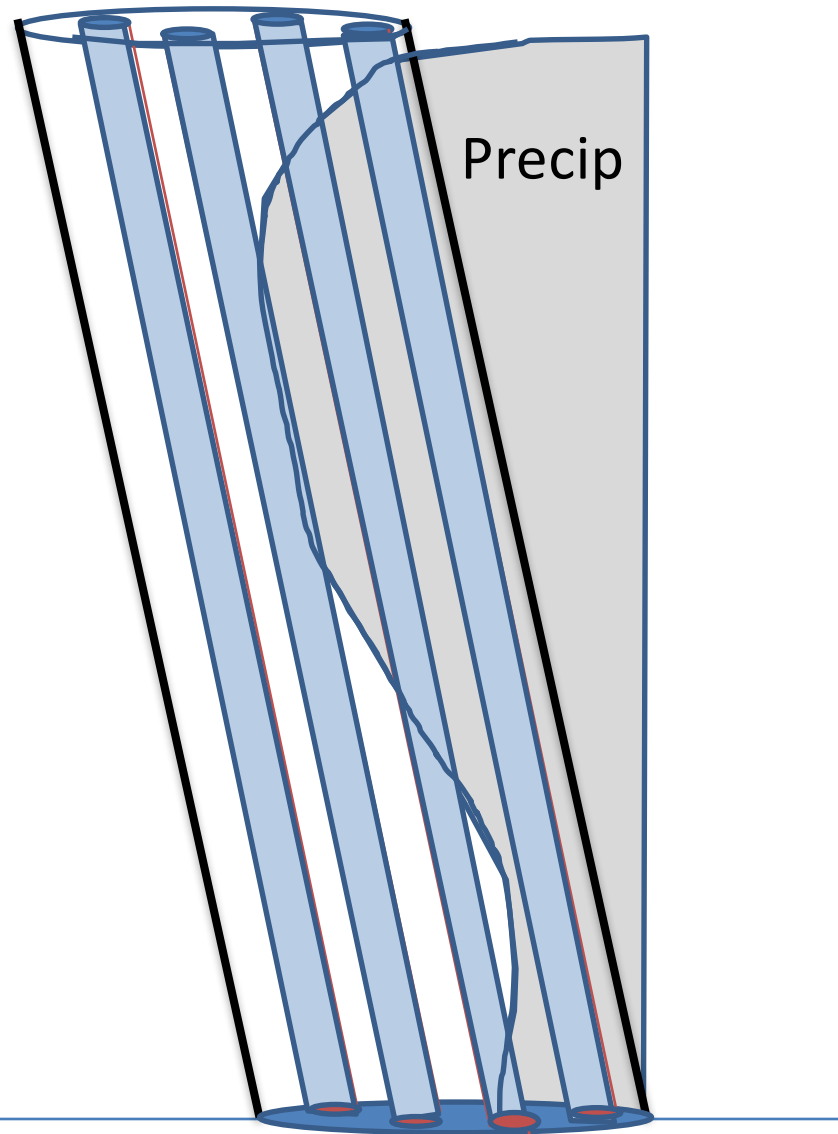
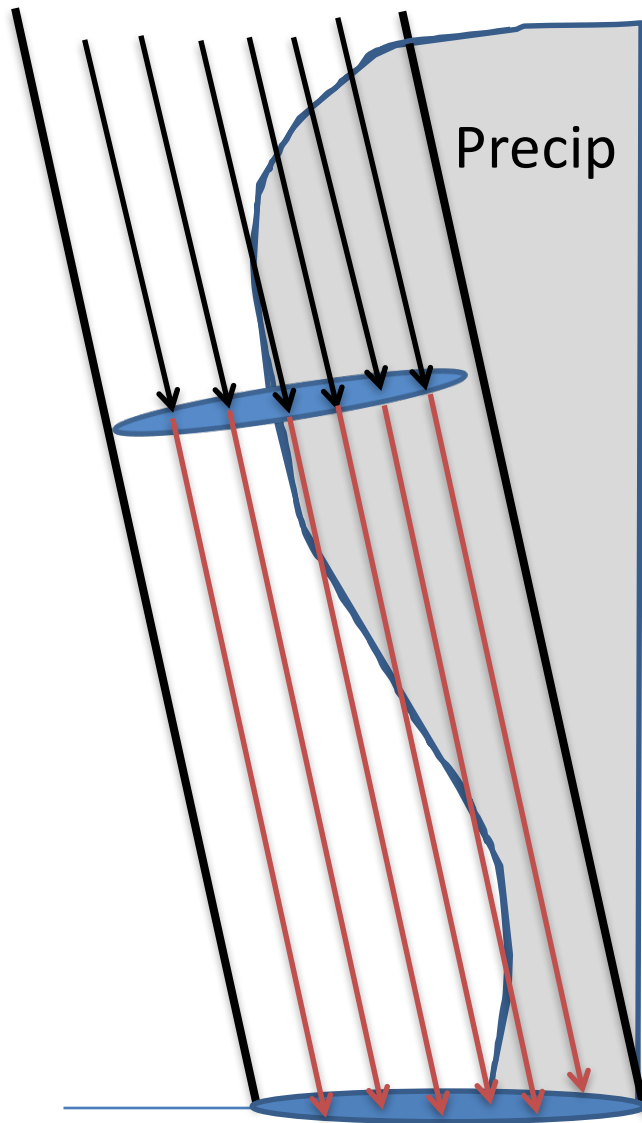






# NUBF Mitigation

- Motivation
  - Attenuation & NUBF are closely linked
  - Attenuation effects exacerbate the NUBF problem
  - As such, the problem is more severe at Ka-band than at Ku-band
  - If we had higher resolution data, the retrieval errors would decrease
- Approach
  - Using ancillary data (in adjacent/interleaved/range-sampled FOV's), interpolate both PIA &  $Z_m$  to higher resolution columns
  - Normalize the interpolated fields to satisfy the initial conditions
  - In this higher resolution space, solve for hi-res  $Z(x,y,z)$  over the multiple columns using traditional methods



# NUBF Mitigation (some eq's)

$$A(\theta) = 2 \int_0^{r_s} k(\theta, s) ds \approx \langle P_s(\text{No Rain}, \theta) \rangle - P_s(\text{Rain}, \theta)$$

$$A(\theta) = -q^{-1} \log \left[ \frac{\iint G^2(x, y) \sigma_R^0(x, y) \exp[-2q \int_0^{r_s} k(s; x, y) ds] dx dy}{\iint G^2(x, y) \sigma_{NR}^0(x, y) dx dy} \right]$$

$$\text{let } g = G^2 / \iint G^2(x, y) dx dy; \quad a(\theta; x, y) = 2 \int_0^{r_s} k(\theta, x, y; s) ds$$

$$\text{assume } \sigma_{NR}^0 = \sigma_R^0; \quad (\text{note that } q = 0.2303)$$

$$A(\theta) = -q^{-1} \log \left[ \iint_{FOV} g(x, y) \exp(-qa(\theta; x, y)) dx dy \right] \quad (1)$$

*if beam is uniformly filled,  $A(\theta) = a(\theta)$*

# NUBF Mitigation

*Similarly*

$$Z_{m,dB}(h) = q^{-1} \log \left[ \iint_{FOV} g(x, y) z_m(x, y; h) dx dy \right] \quad (2)$$

*Note that near the surface, the hi – res fields are related by*

$$z(x, y) = z_m(x, y) \exp(-qa(x, y))$$

*where  $z(x, y)$  is the hi – res, attenuation – corrected reflectivity factor*

# NUBF Mitigation

*Replace high-res fields with the interpolated fields along with adjustment factors*

$$a(x, y) \rightarrow \tilde{a}(x, y) + \delta_a$$

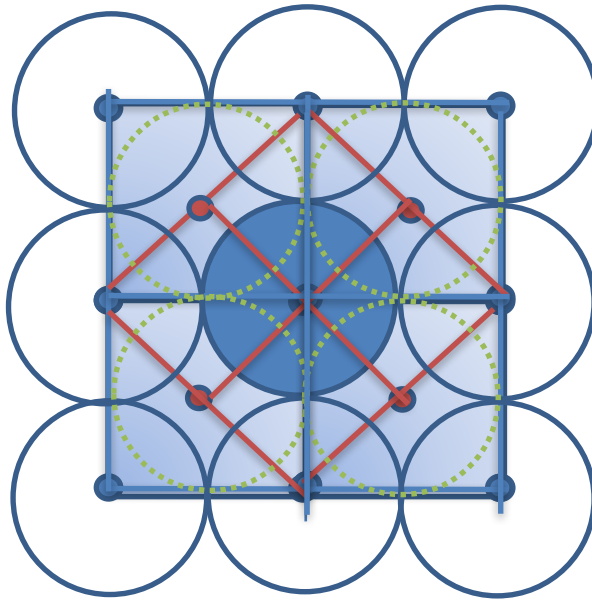
$$z_m(x, y) \rightarrow \delta_{z_m} \tilde{z}_m(x, y)$$

*adjust  $\delta_a$  and  $\delta_{z_m}$  so that (1) and (2) are satisfied*

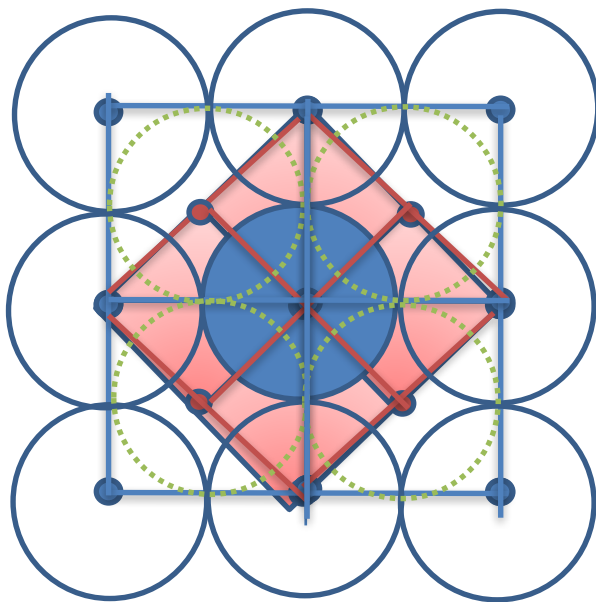
*Use modified interpolated fields in standard retrieval equations to get  $Z(x, y, z)$  at interpolated resolution*



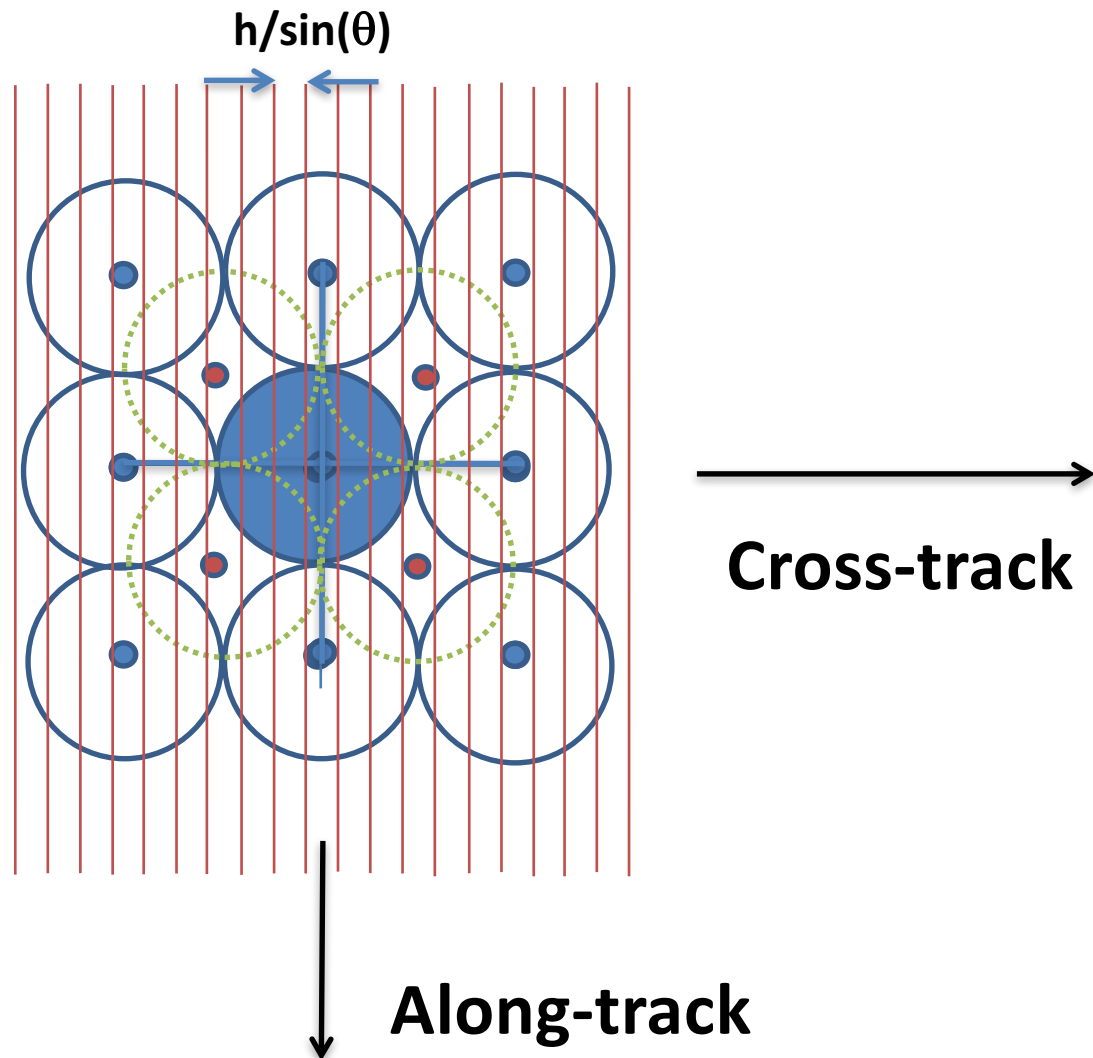
## Nadir Geometry (Standard Data)



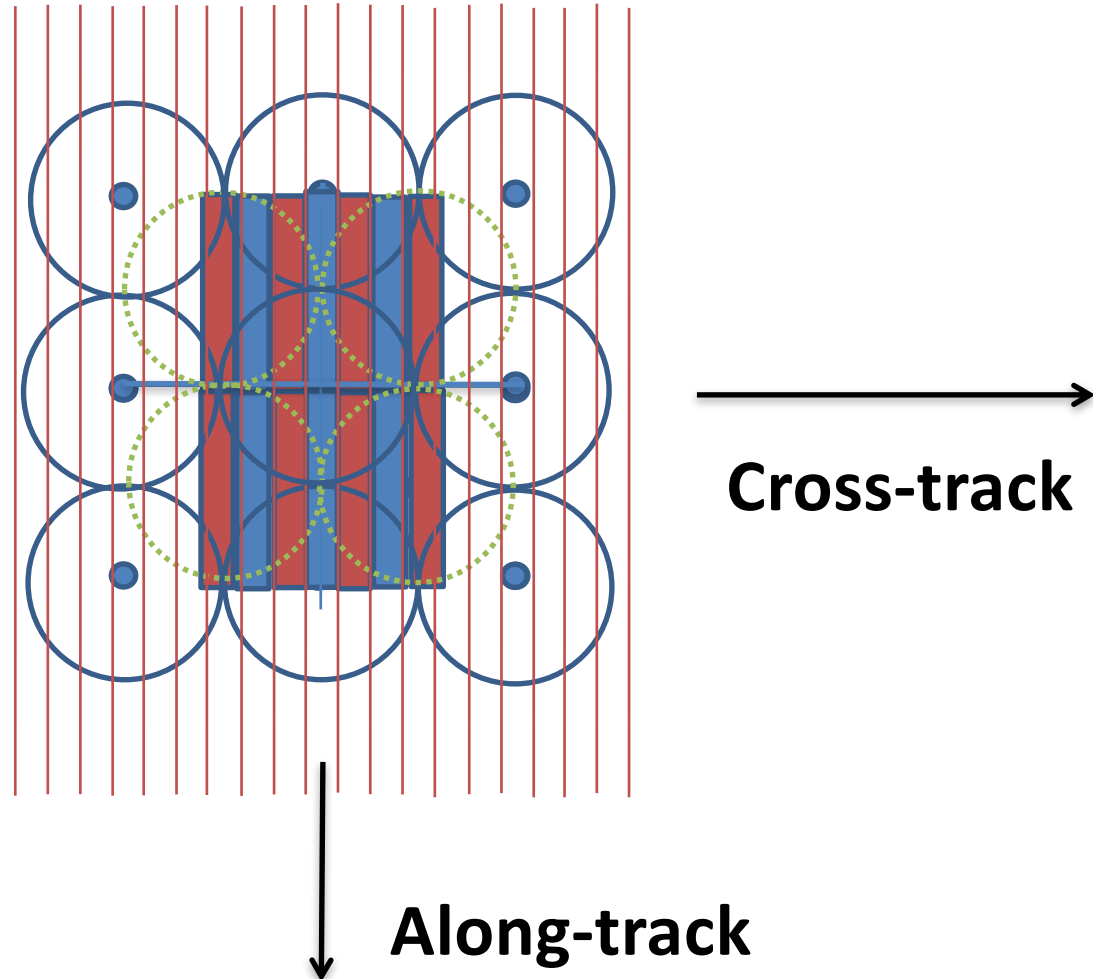
## Nadir Geometry (w/ Interleaved Data)



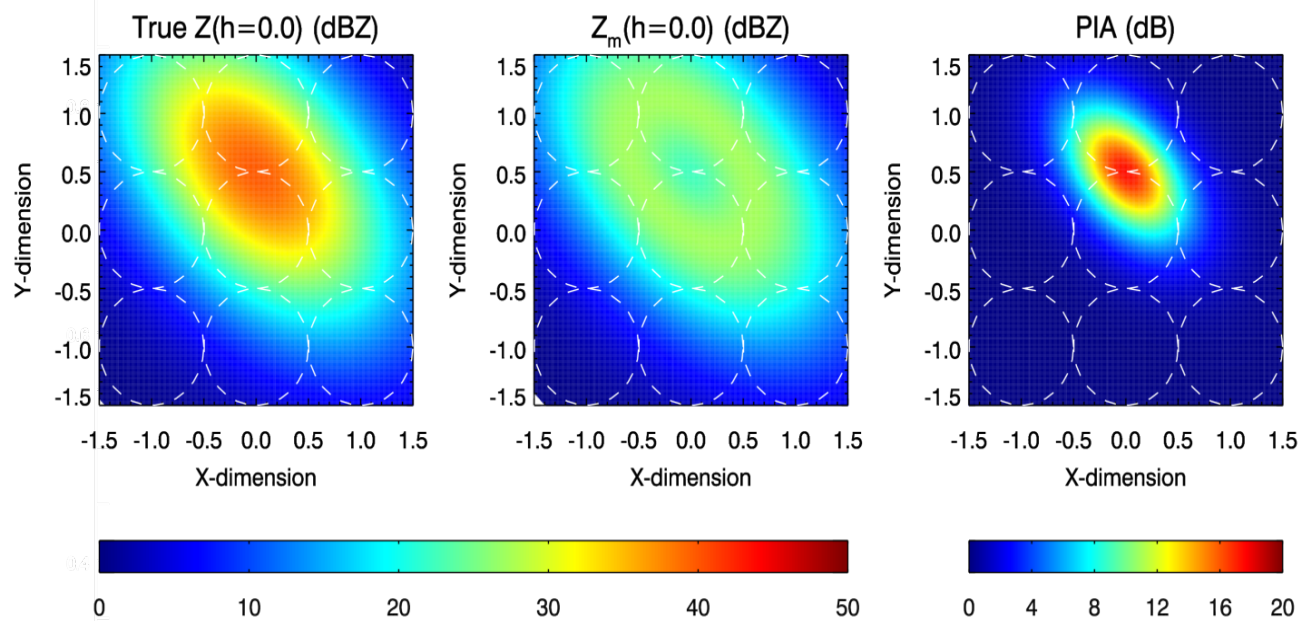
# Off-Nadir Geometry with Range-Profiled PIA

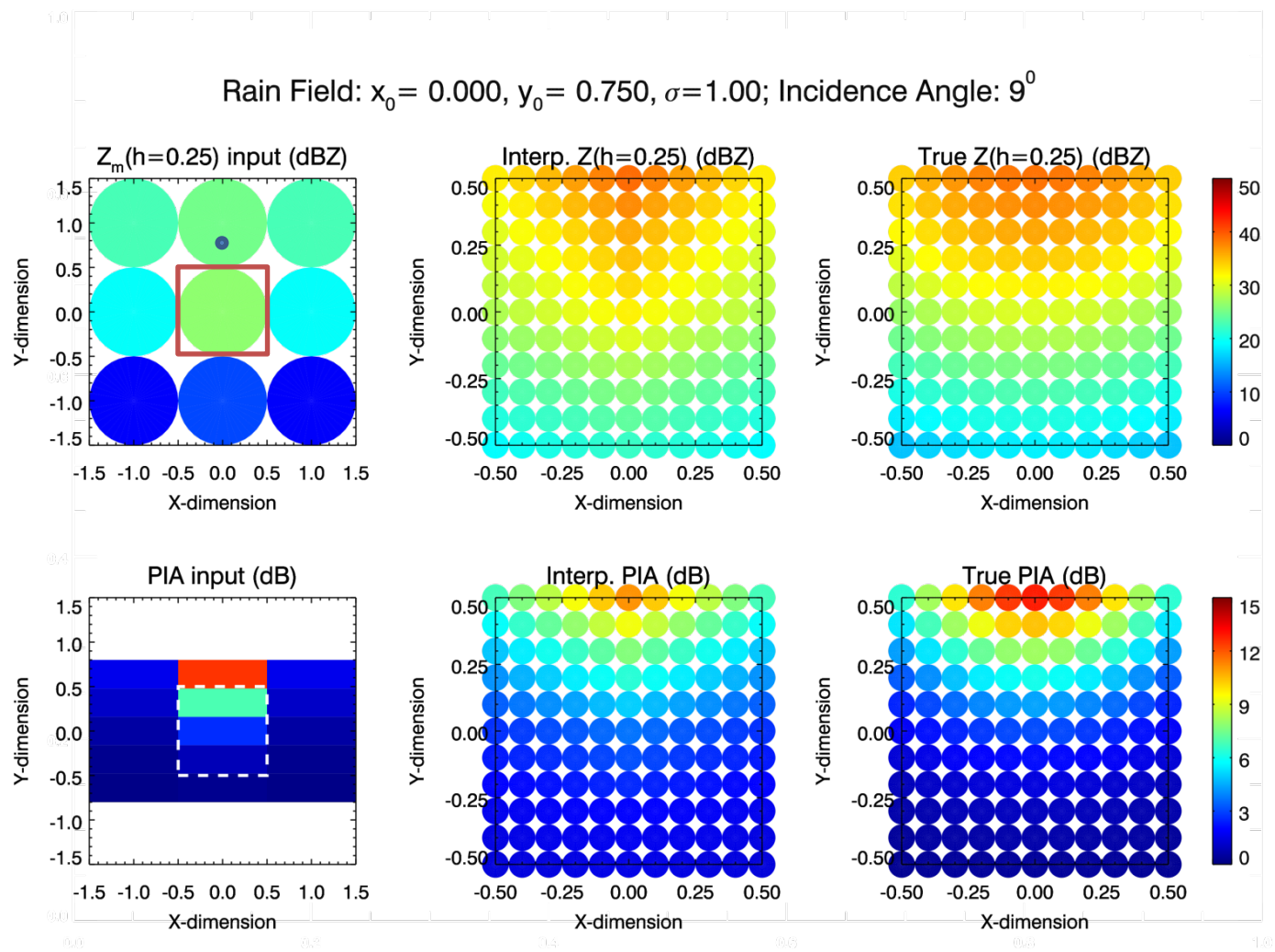


# Off-Nadir Geometry with Range-Profiled PIA

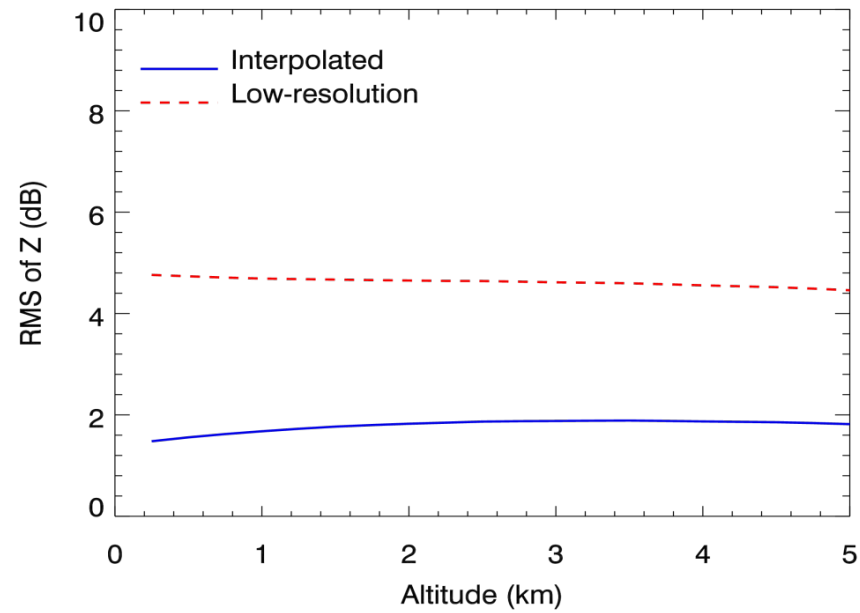
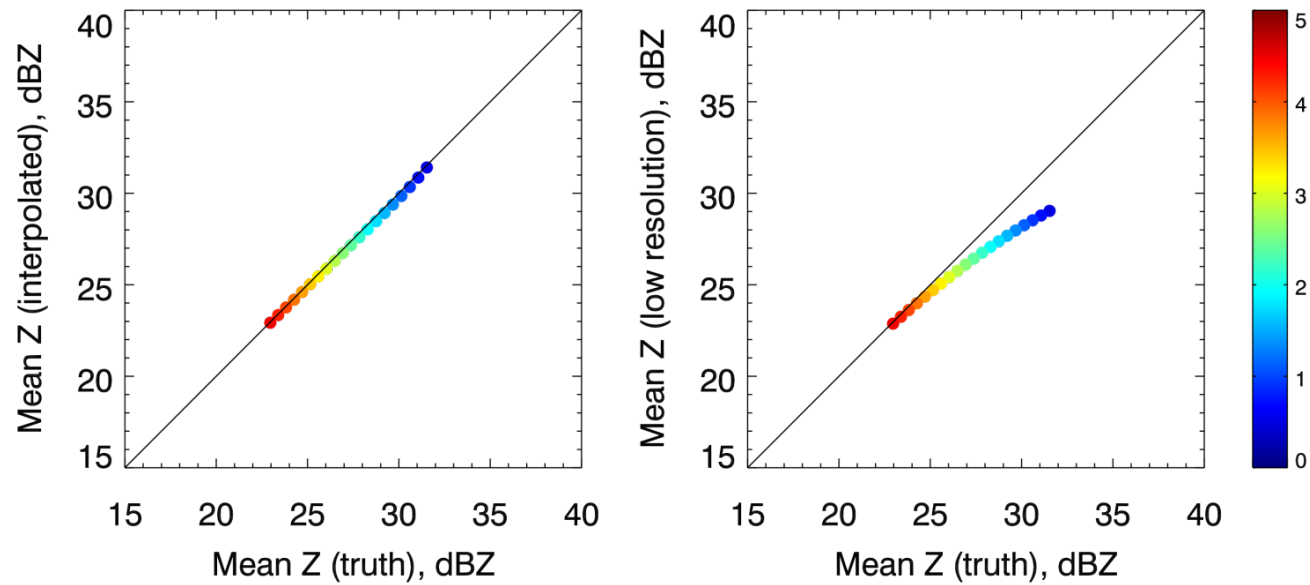


Rain Field:  $x_0 = 0.000$ ,  $y_0 = 0.500$ ,  $\sigma = 1.00$ ; Incidence Angle:  $0^\circ$

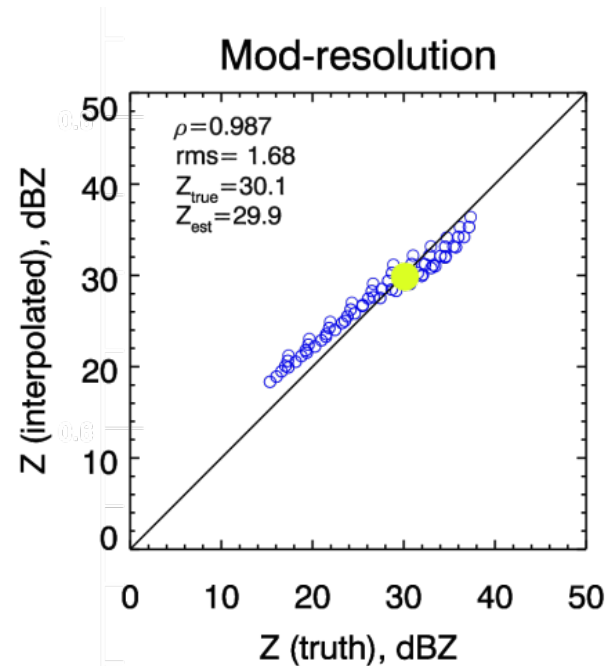
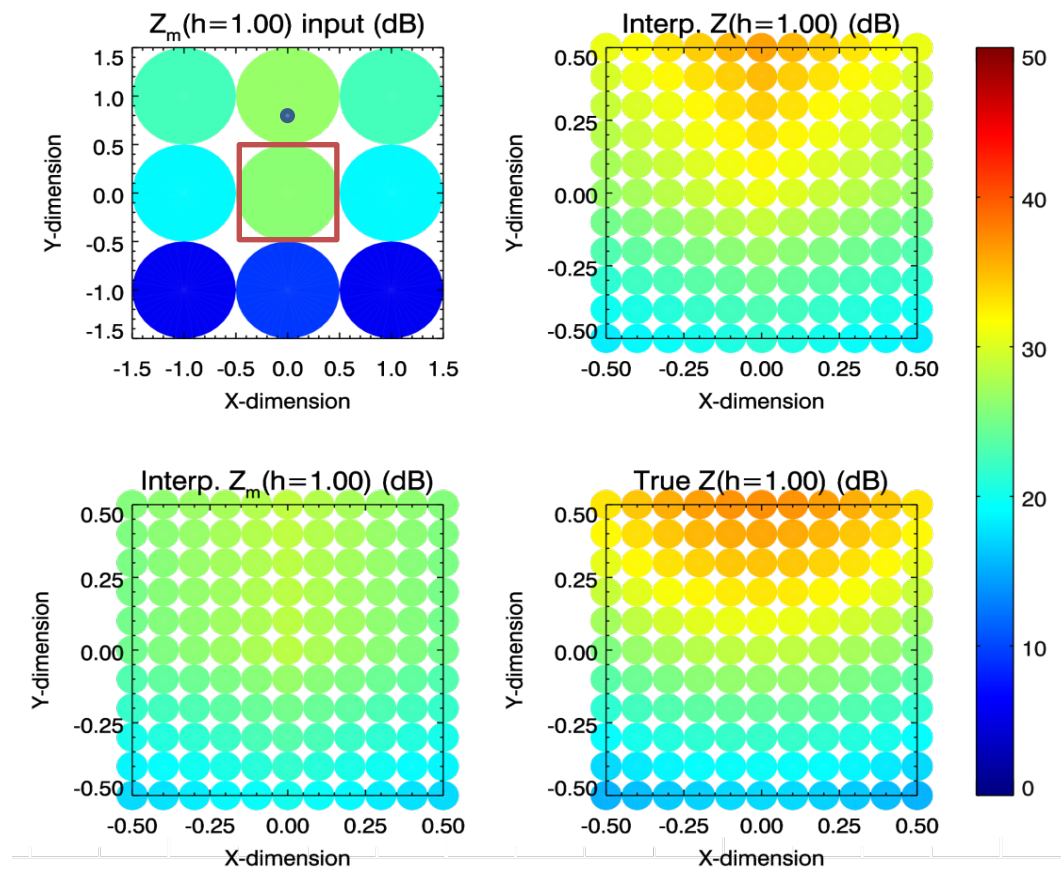




Rain Field:  $x_0 = 0.000$ ,  $y_0 = 0.750$ ,  $\sigma = 1.00$ ; Incidence Angle:  $9^\circ$

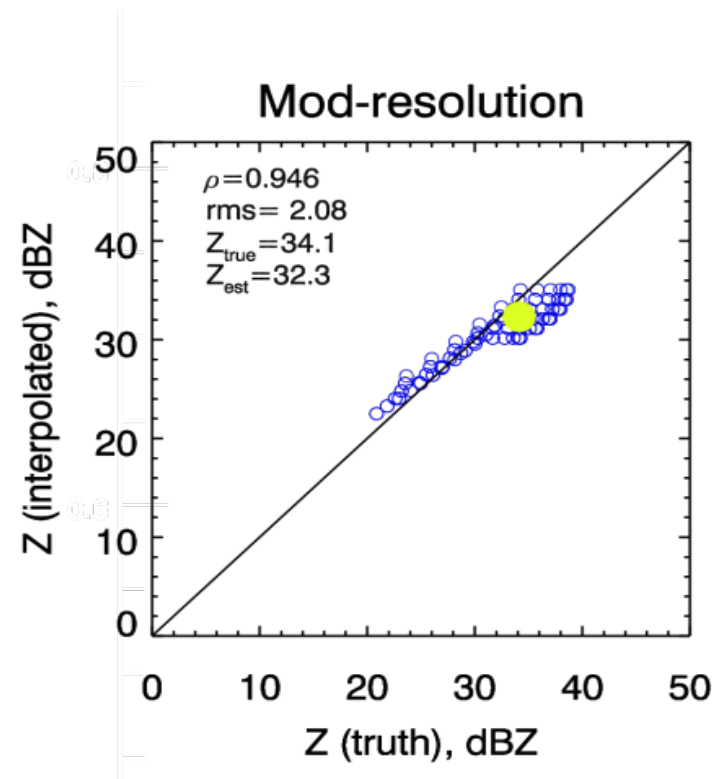
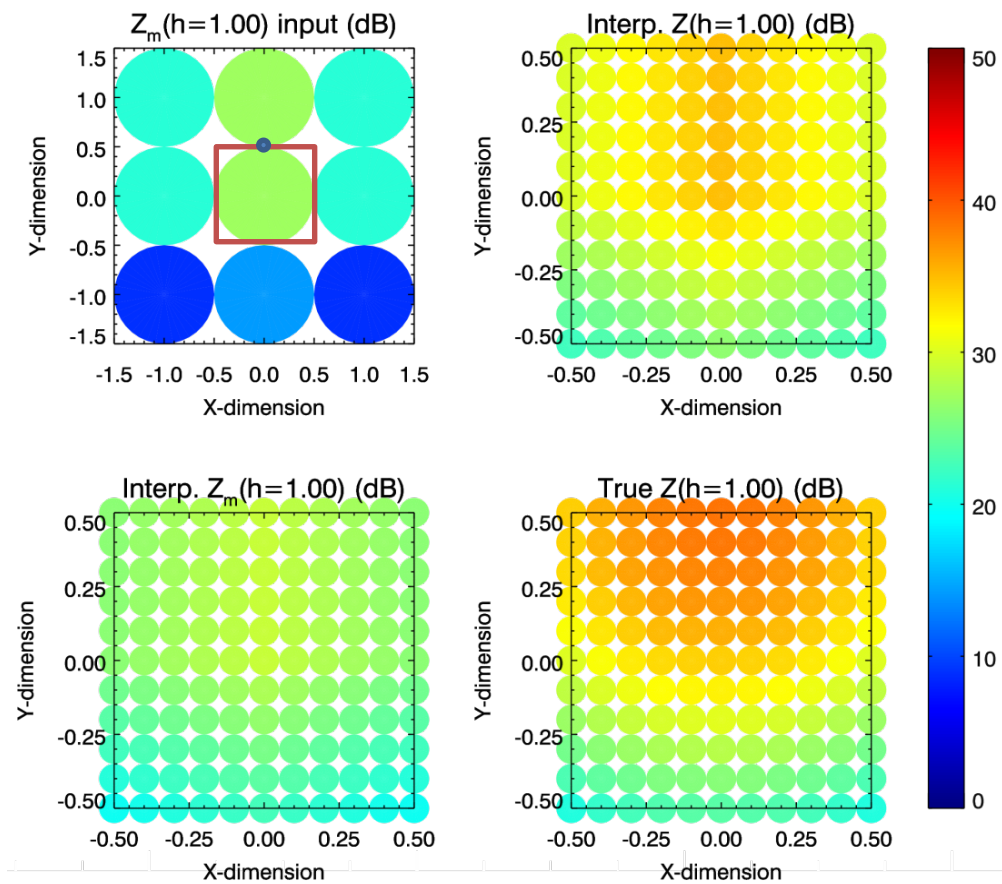


Rain Field:  $x_0 = 0.000$ ,  $y_0 = 0.750$ ,  $\sigma = 1.00$ ; Incidence Angle:  $9^\circ$

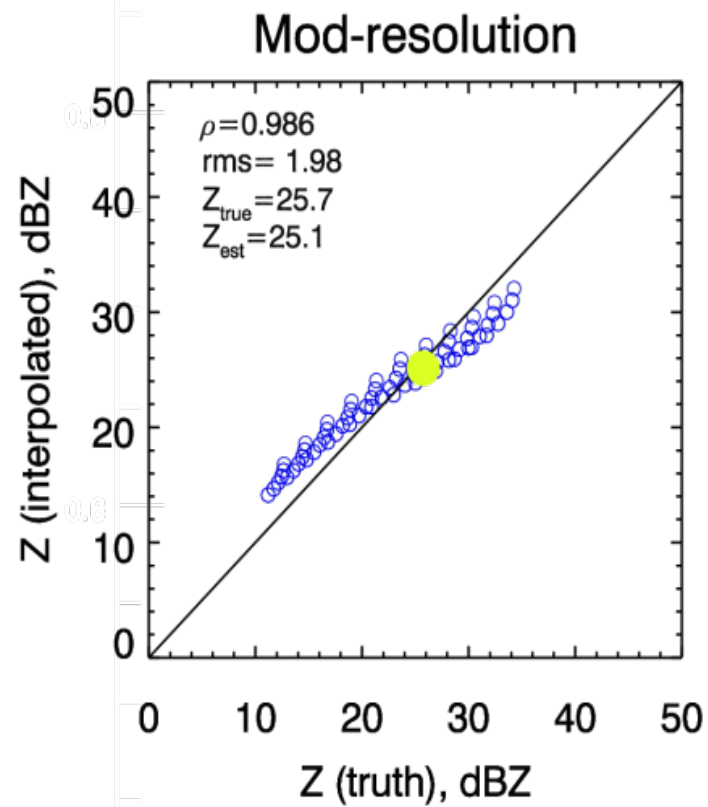
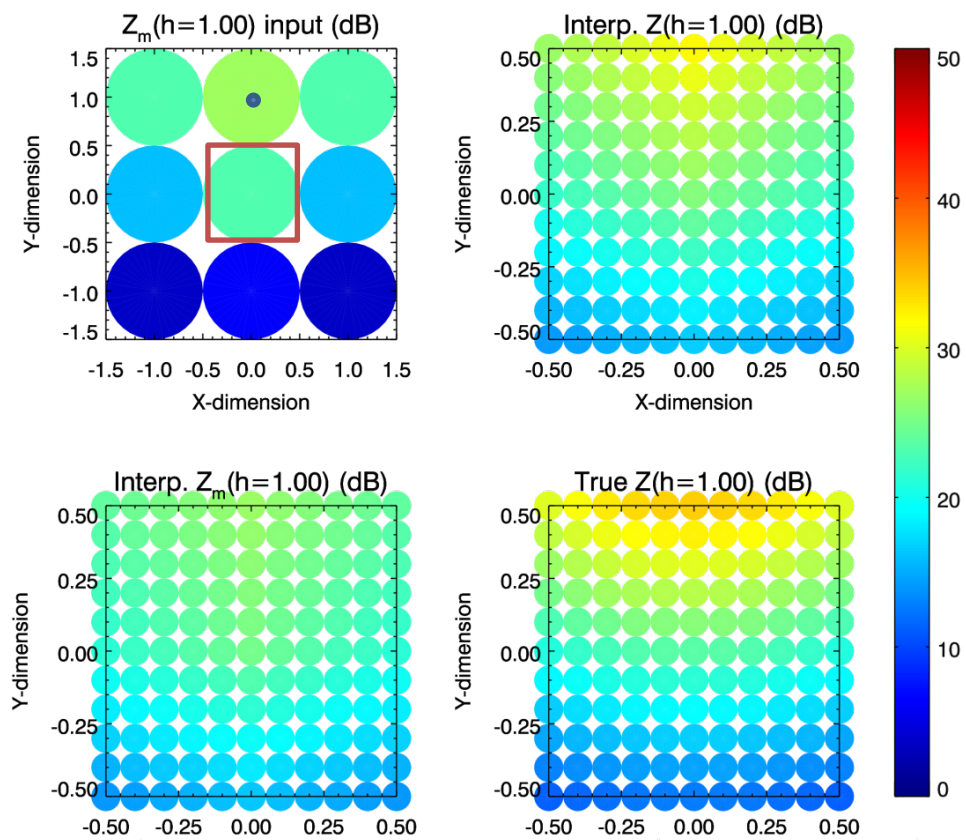




Rain Field:  $x_0 = 0.000$ ,  $y_0 = 0.500$ ,  $\sigma = 1.00$ ; Incidence Angle:  $0^\circ$



Rain Field:  $x_0 = 0.000$ ,  $y_0 = 1.000$ ,  $\sigma = 1.00$ ; Incidence Angle:  $0^\circ$



# Comments

- The procedure gives some improvement, usually modest, over coarse resolution estimate
  - Greater reduction in rms error than in bias
  - Degree of improvement is non-uniform, however
- Bilinear interpolation has been used
  - Kriging & other geospatial methods might provide better results, esp when using interl. Ka-band data
- To understand the method, a simple storm model is used
  - MRMS data are being used to get a more realistic assessment of the approach

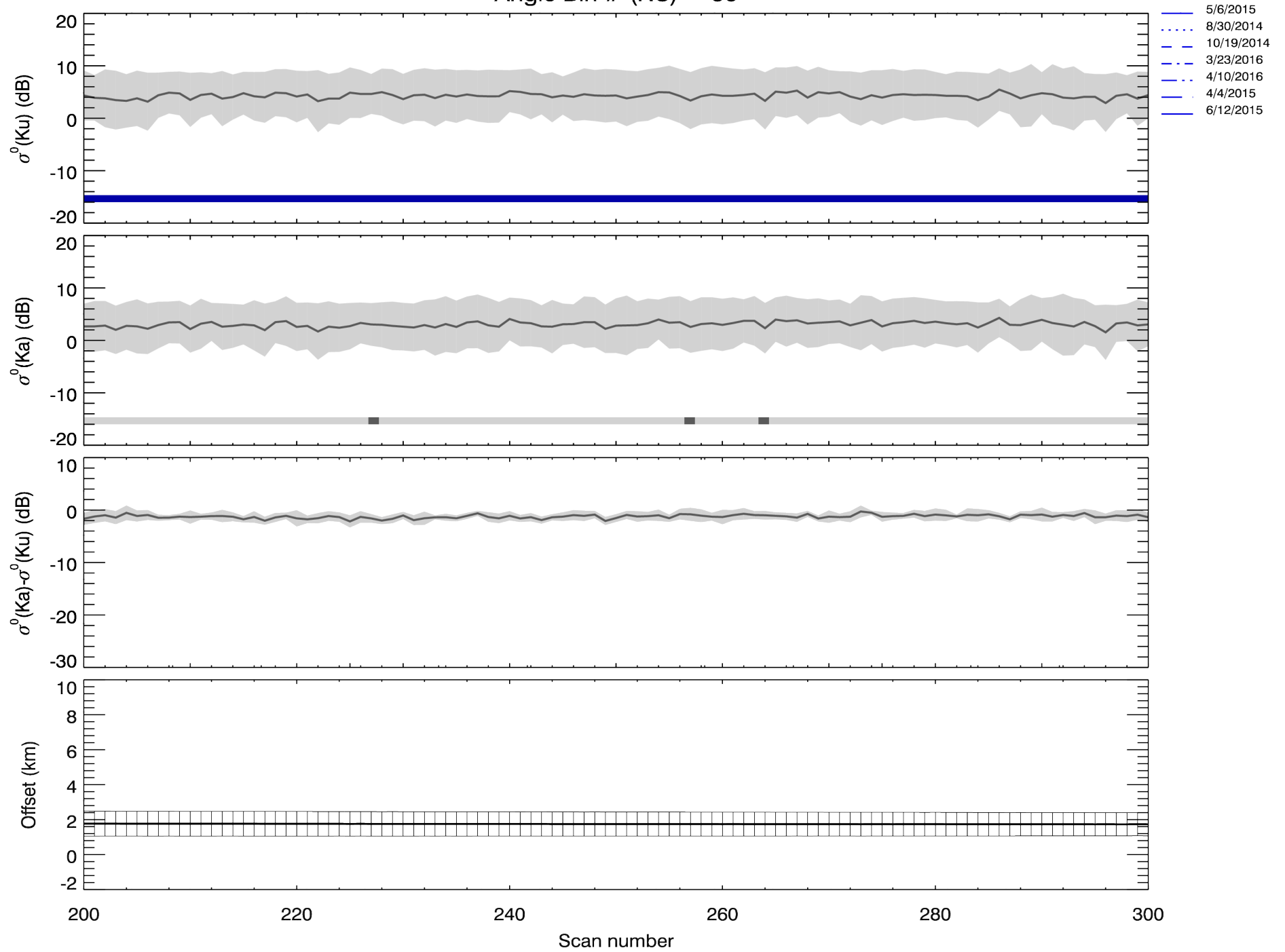
# Summary

- Changes in SRT code & data bases have been made for V5
- Several improvements in the method appear to be feasible
  - Correction for under-sampling surface power at nadir
  - Variable spatial averaging of temporal reference data
  - Implementation of 5 surface categories
- An NUBF-mitigation strategy has shown some potential & will be pursued

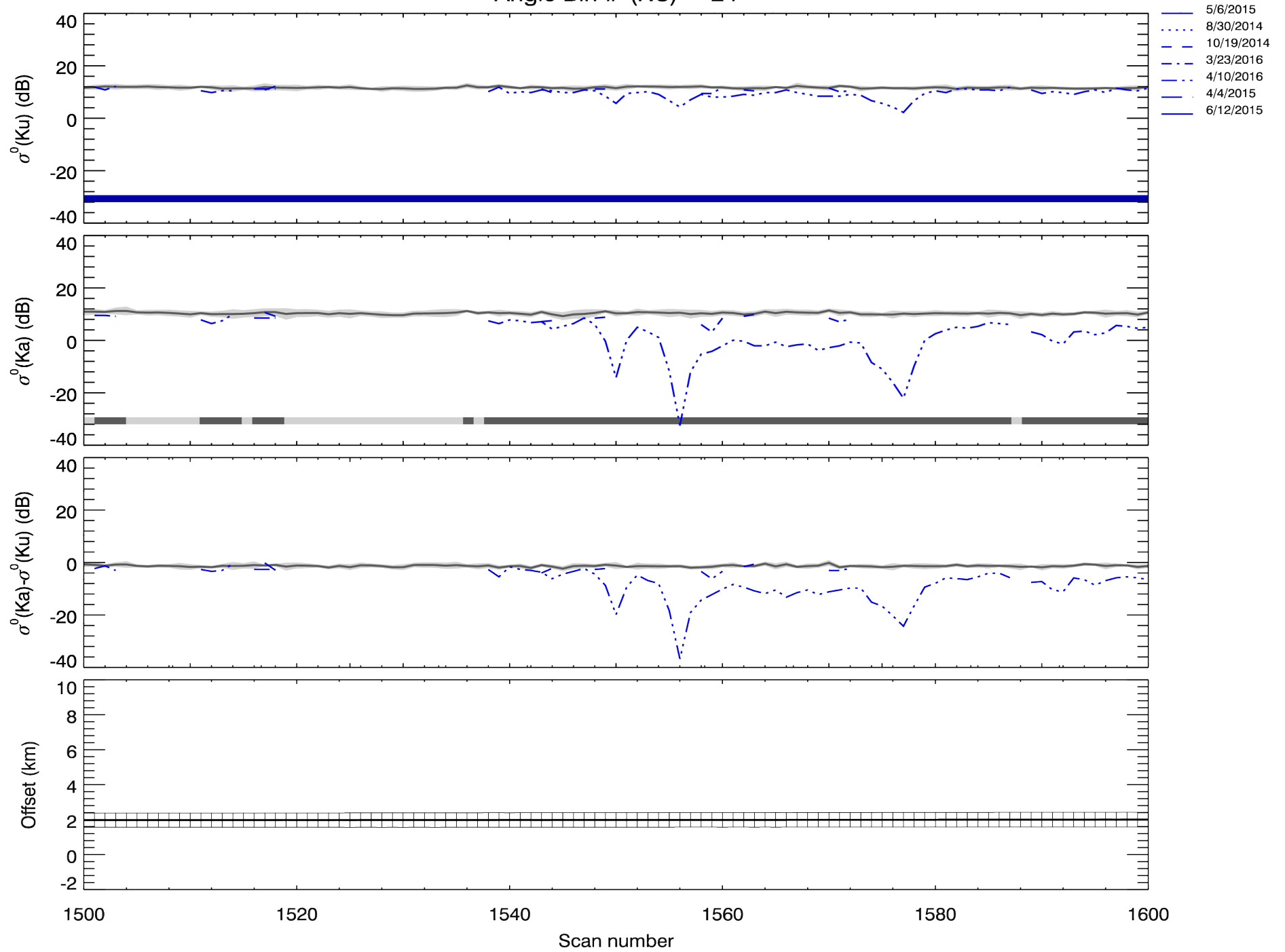
# PIA Estimates from Temporal Reference Data

- In following slides, the mean & std dev of the rain-free  $\sigma^0$  data are shown by the black line and gray area about the black line ( $\pm 1$  std dev)
- Surface type is indicated by the bar in top panel
- Rain/no-rain is indicated by the bar in the 2<sup>nd</sup> panel
- $\sigma^0$  data in rain are depicted by the blue lines
  - Different line types represent different orbits
- Difference between black & blue lines gives a temporal-ref estimate of the 2-way PIA
- The error assoc with the PIA is proportional to the vertical extent of the gray area

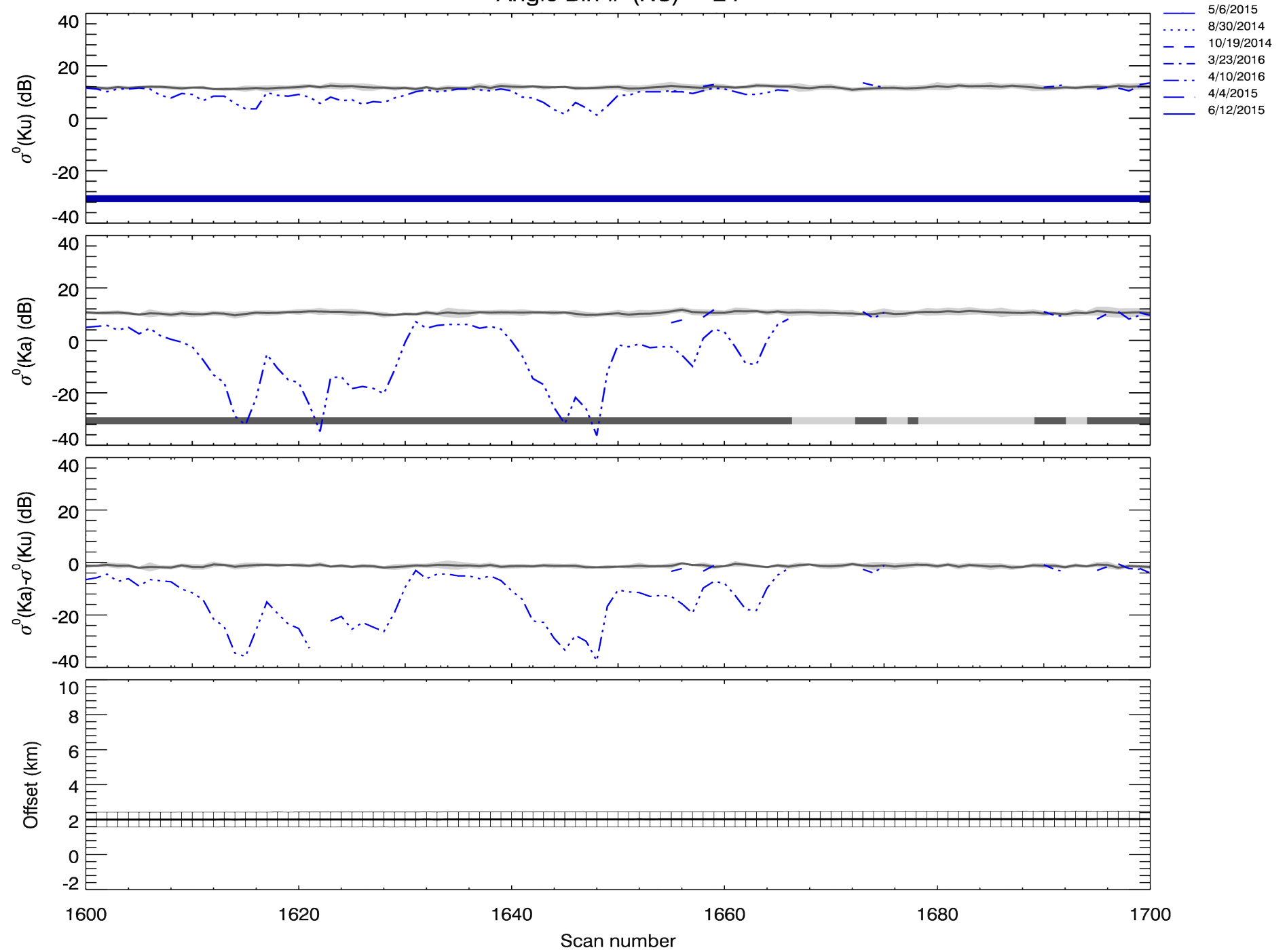
Angle Bin # (NS) = 36



Angle Bin # (NS) = 24

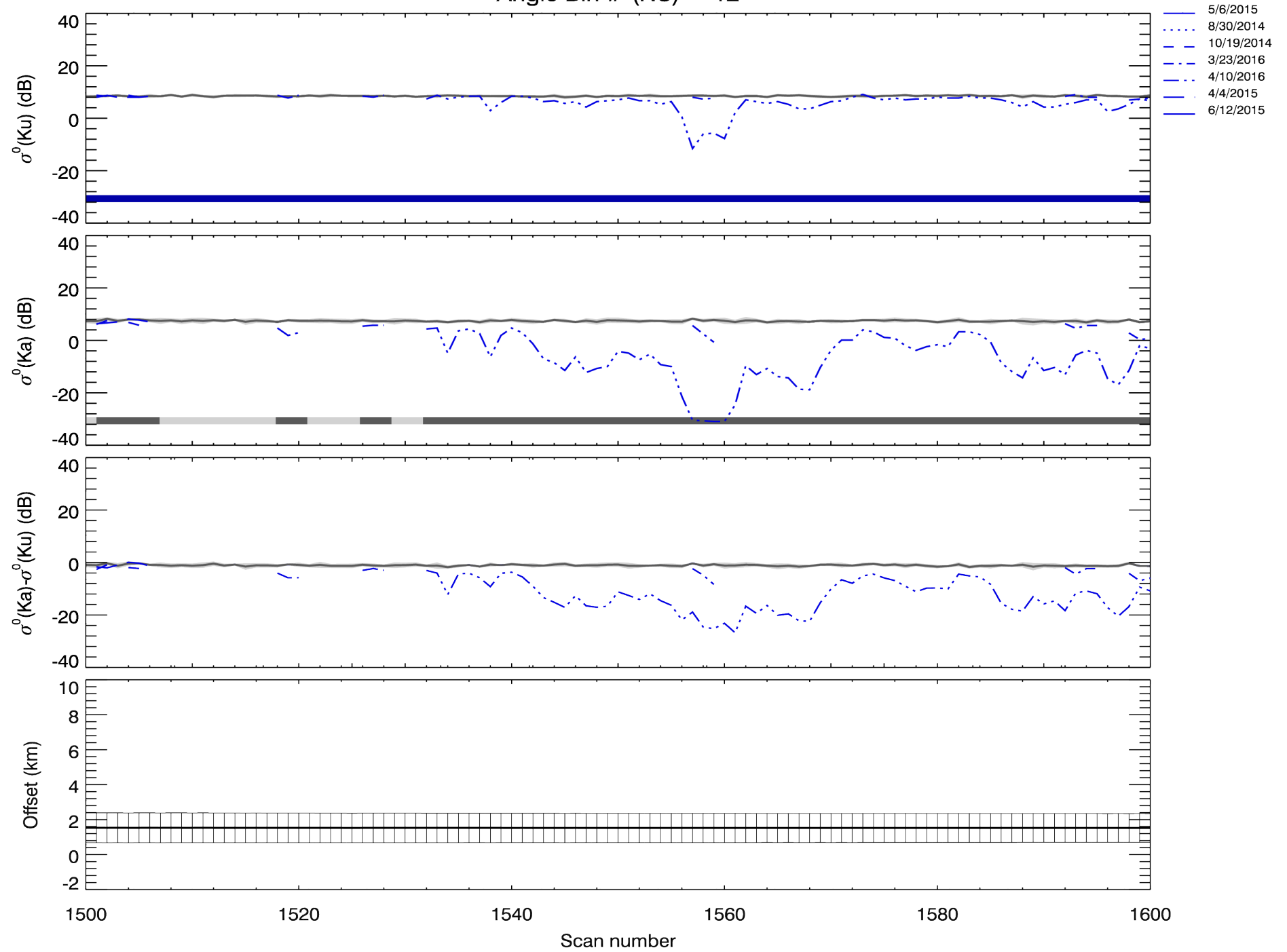


Angle Bin # (NS) = 24





Angle Bin # (NS) = 12



Angle Bin # (NS) = 24

